



# Predicción climática decadal: experiencia y perspectivas

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Earth Sciences Dept. (BSC-ES)



based on our work during 2010-2016 at IC3 + BSC-ES

# Predicción climática decadal: experiencia y perspectivas



**Climate  
Forecasting  
Unit**

2010-2015



2015-2016

COMPUTATIONAL EARTH SCIENCES

ATMOSPHERIC  
COMPOSITION

CLIMATE  
PREDICTION

EARTH SYSTEM SERVICES

## Multi-model decadal forecast exchange

The Met Office coordinates an informal exchange of near-real time decadal predictions. Many institutions around the world are developing decadal prediction capability and this informal exchange is intended to facilitate research and collaboration on the topic.

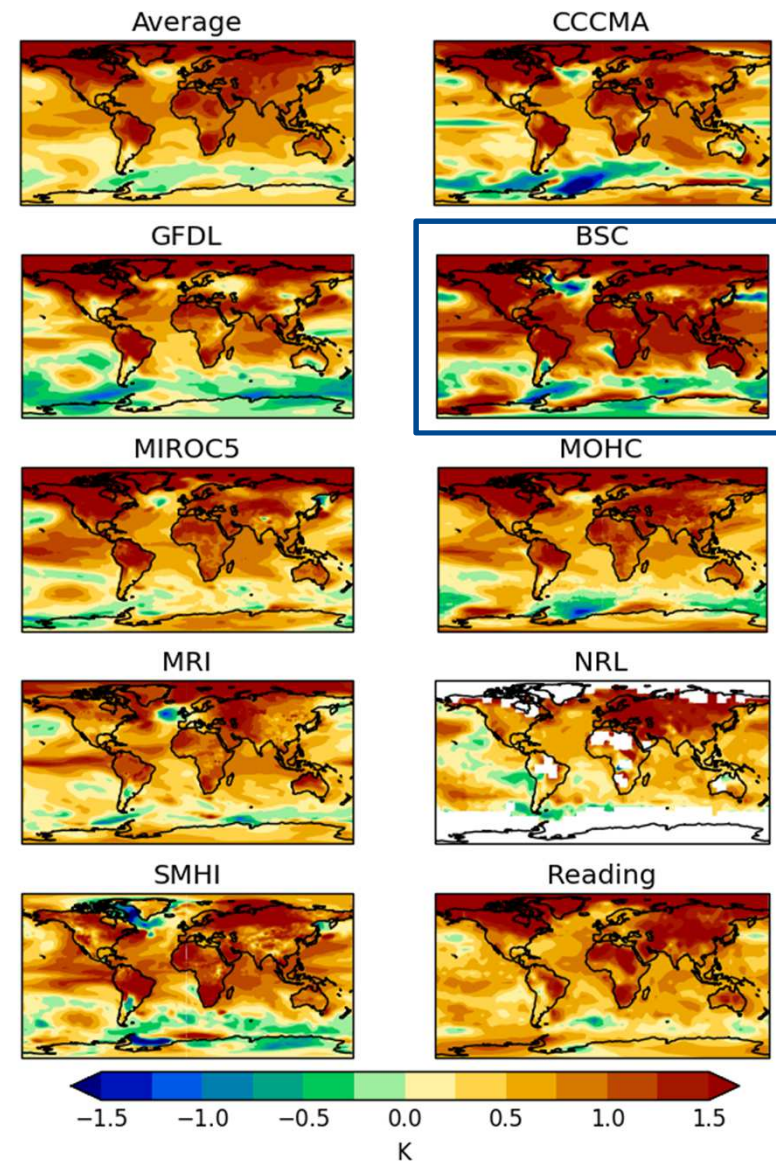
[The contributing prediction systems](#) are a mixture of dynamical and statistical methods. The prediction from each institute is shown below, alongside an average of all the models. When possible, observations for the period of the forecast are also shown. Currently three variables are included: surface air temperature, sea-level pressure and precipitation. These are shown as differences from the 1971-2000 baseline. More diagnostics, including ocean variables are planned for the future. Please use the drop-down menus below to explore the data collected to date.

This work is supported by the European Commission SPECS project.

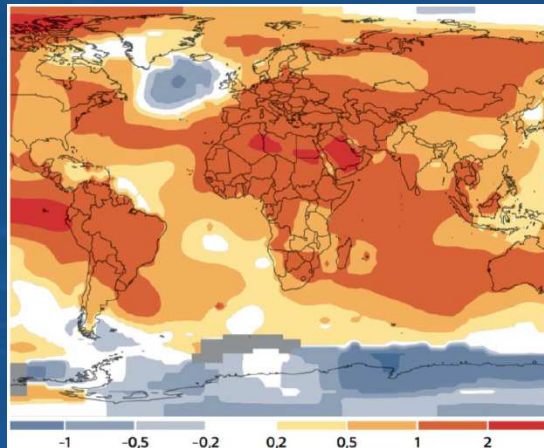


Smith et al. (2013, ClimDyn) - UPDATED

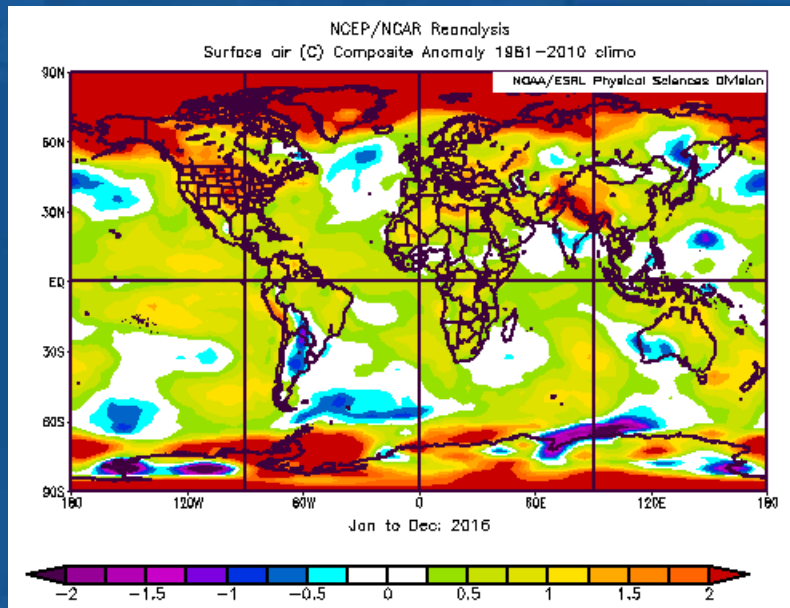
2015 predictions for 2016 surface temperature



May-Oct 2015

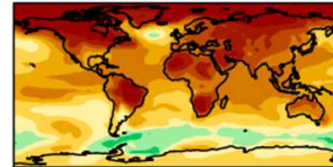


Jan-Dec 2016

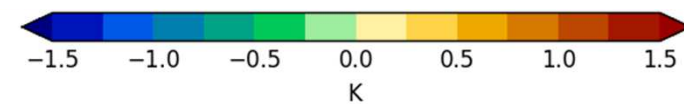
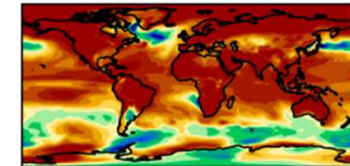


2015 predictions for 2016 surface temperature

Average

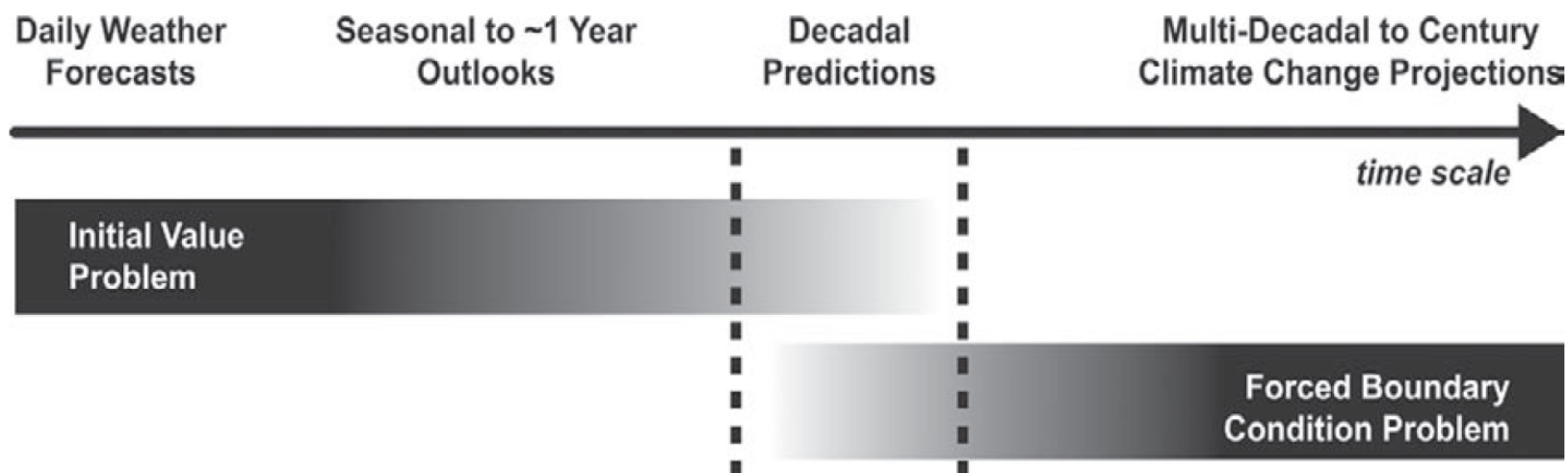


BSC

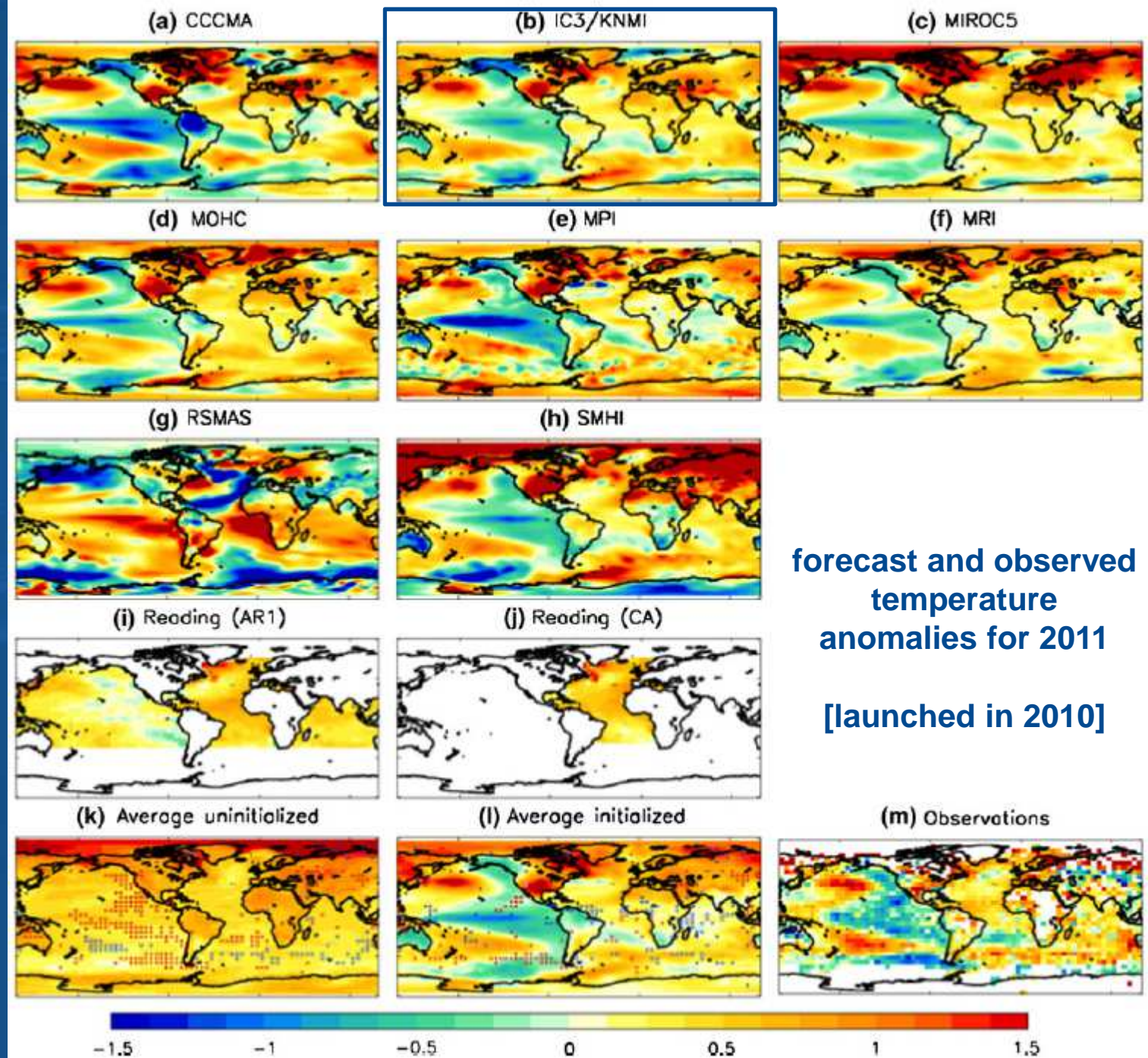




- Decadal prediction bridges seasonal forecasting and climate-change projection
- The decadal timescale is a key planning horizon for adaptation/decision making
- Decadal prediction not only deals with anthropogenically forced variability, but also with natural (unforced) internal variability
- Decadal prediction opens the possibility of achieving skill in multi-year timescales, and in both regional and large-scale domains



Meehl et al. (2009, BAMS)



**forecast and observed  
temperature  
anomalies for 2011**

**[launched in 2010]**

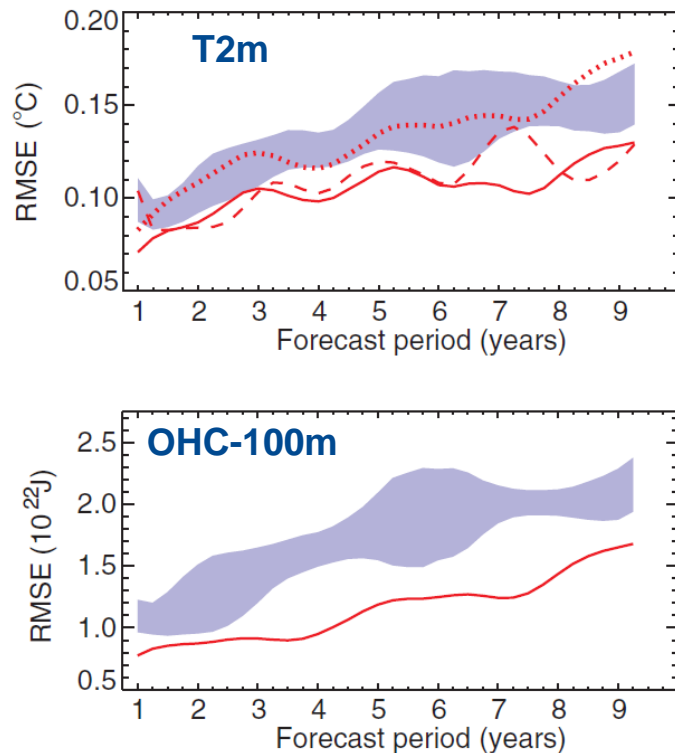
# Decadal climate prediction



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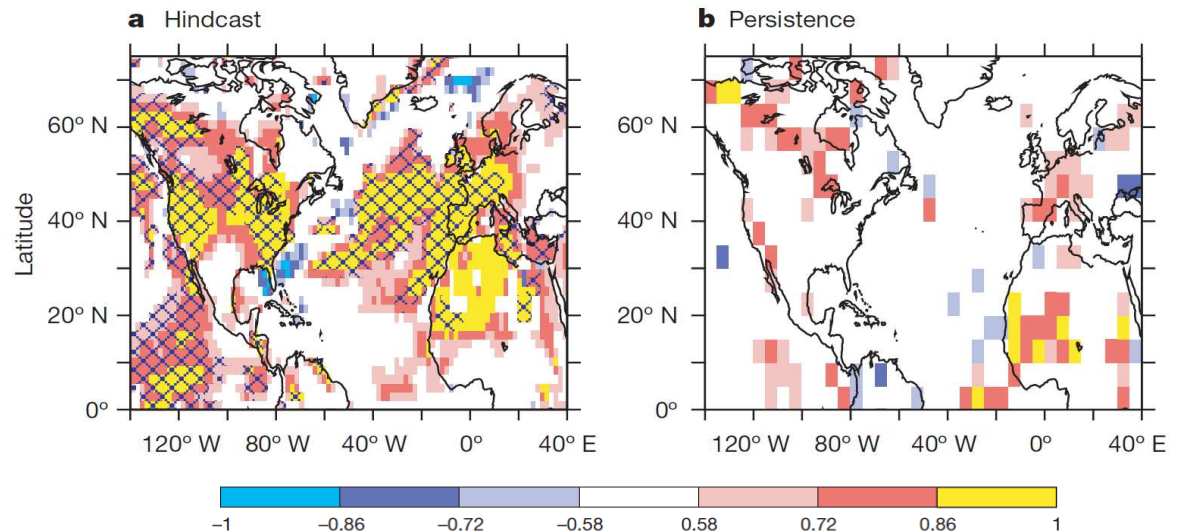
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## DePreSys (UKMO)

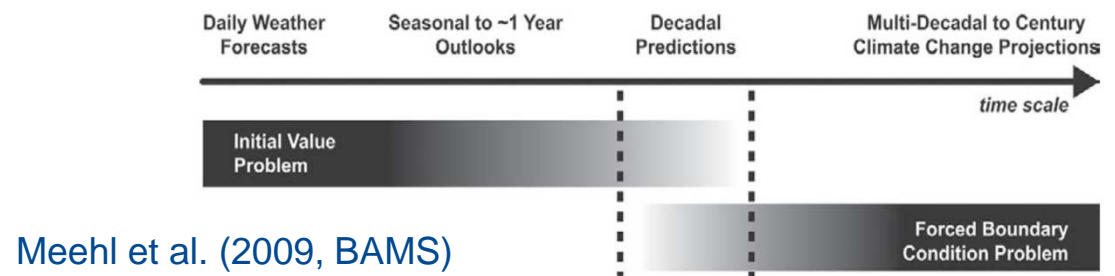


Smith et al. (2007, Science)

## ECHAM5/MPI-OM (GEOMAR & MPI)



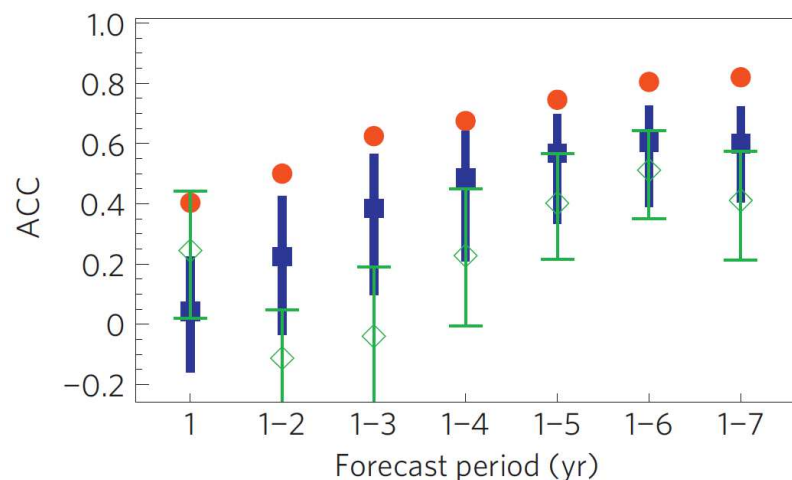
Keenlyside et al. (2008, Nature)



Meehl et al. (2009, BAMS)



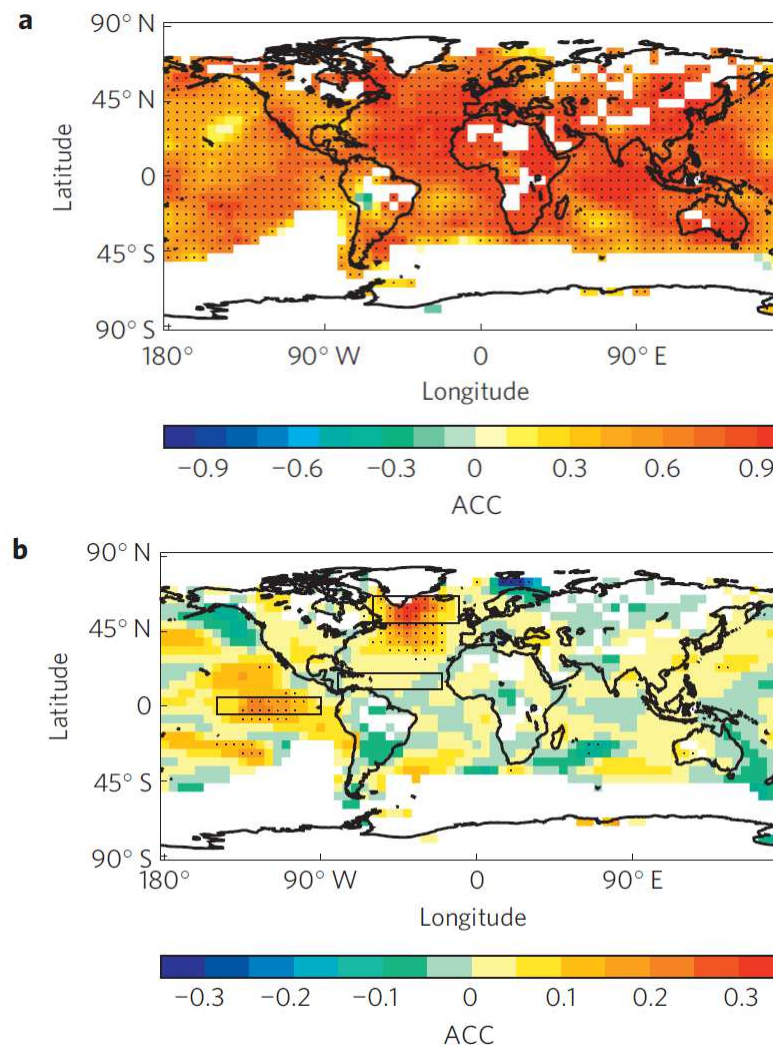
## DePreSys (UKMO)



**Figure 2 | Multi-annual hindcasts of Atlantic tropical storm frequency.**

**a**, ACC (see Supplementary Information) for predictions of the number of Atlantic tropical storms for increasing forecast periods. Forecast period '1' is the first hurricane season (months 8-13 from November hindcasts), and '1-7' is the average of years 1-7 inclusive. Initialized predictions (DePreSys, red circles) are compared with externally forced (NoAssim, blue squares) and persistence (green diamonds, see the Methods section), with the blue/green bars indicating the 5-95% confidence interval in which differences in skill from DePreSys are not significant (see Supplementary Information).

Smith et al. (2010, Nature Geoscience)



ACC between 5-year mean (Jun-Nov) surface temperature predicted by DePreSys-Assim and HadCRUT3 (a); ACC difference between Assim and NoAssim (b)

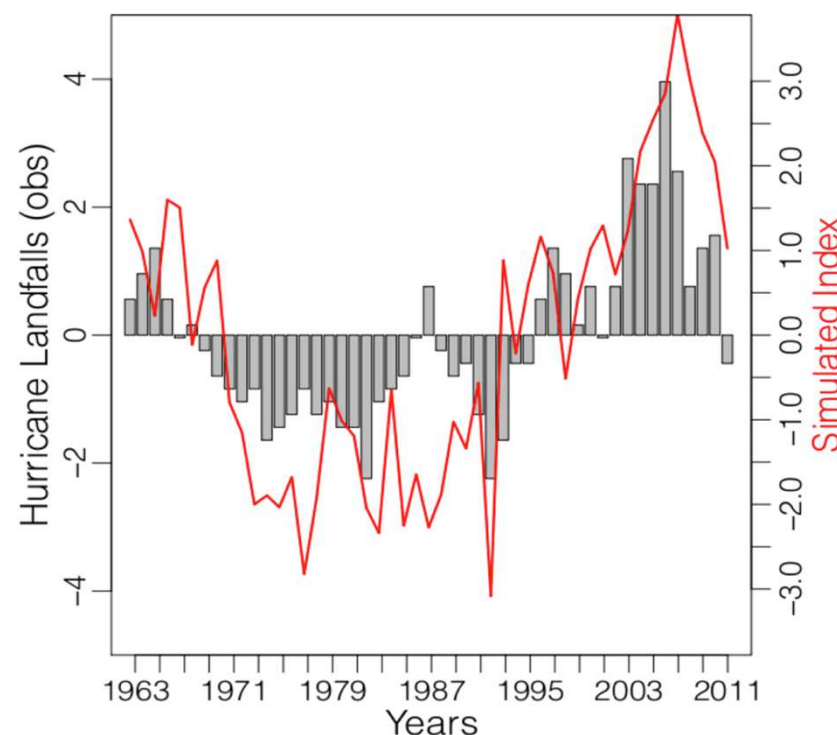
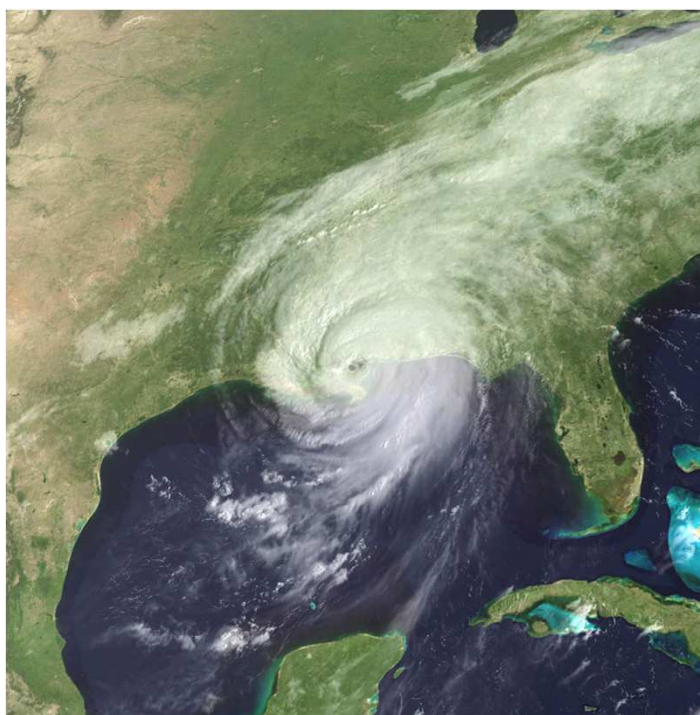


AMO is a predictor for tropical cyclone activity in the Atlantic.  
Decadal predictions over forecast years 1-5, including the  
Accumulated Cyclone Energy (ACE) diagnostic.

XL

XL CATLIN  
- INSURANCE

XL CATLIN  
- REINSURANCE



Camp and Caron (2016, Springer)

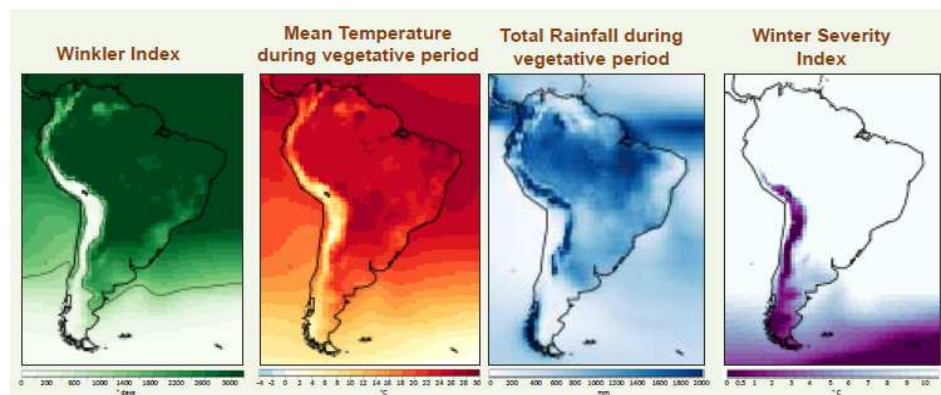
## SECTEUR:

Sector Engagement for the Copernicus Climate Change Service (C3S):  
Translating European User Requirements

Bodegas Torres (a Spanish winery) is looking for new locations for its vineyards (and it's not the only one doing it).

Land is being purchased closer to the Pyrenees, at higher elevation. They are considering acquiring land in South America too, in areas where wine is currently not produced.

Bodegas Torres requests local climate information (including appropriate uncertainty assessments) for the vegetative cycle of the vine.



VIÑAS  
DEL VERO



SOGRAPE VINHOS

COGNAC  
J. & F. COGNAC



RAIMAT



1870



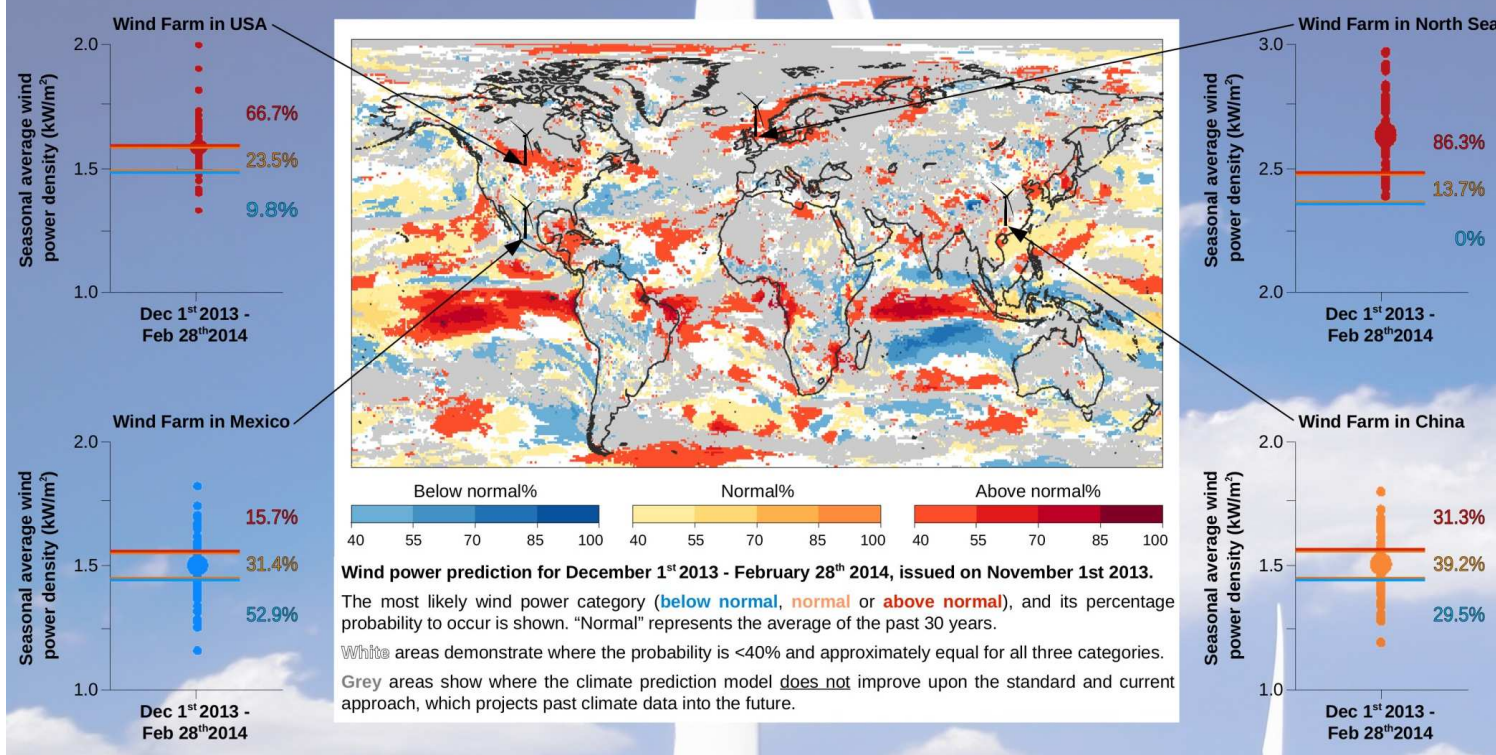
## WIND POWER PREDICTION



renewables



### Illustrative examples of seasonal wind power predictions



at decadal time-scale:  
(e.g. capacity factor ~T2m and 10m.wind)

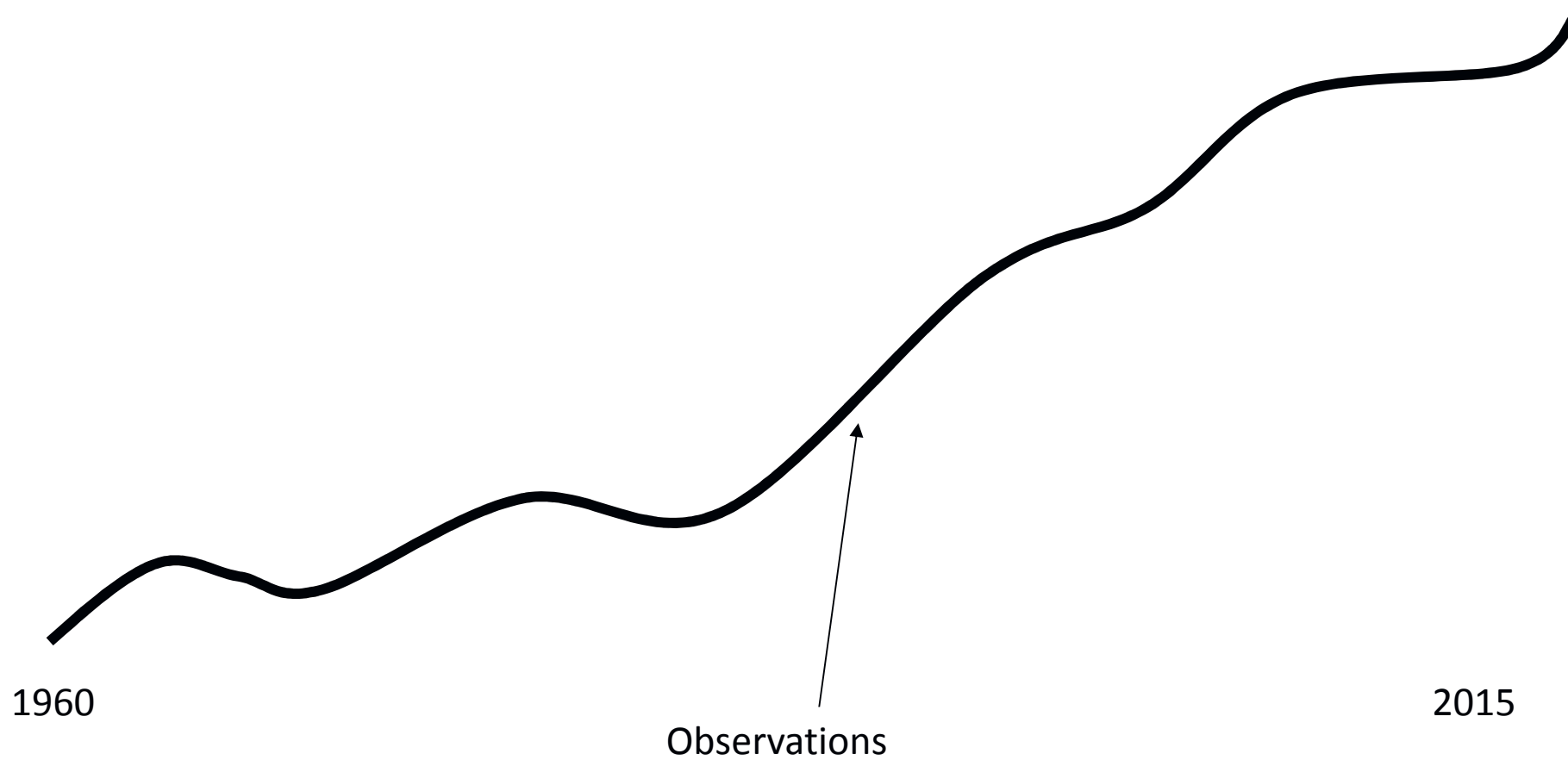




# Decadal climate prediction



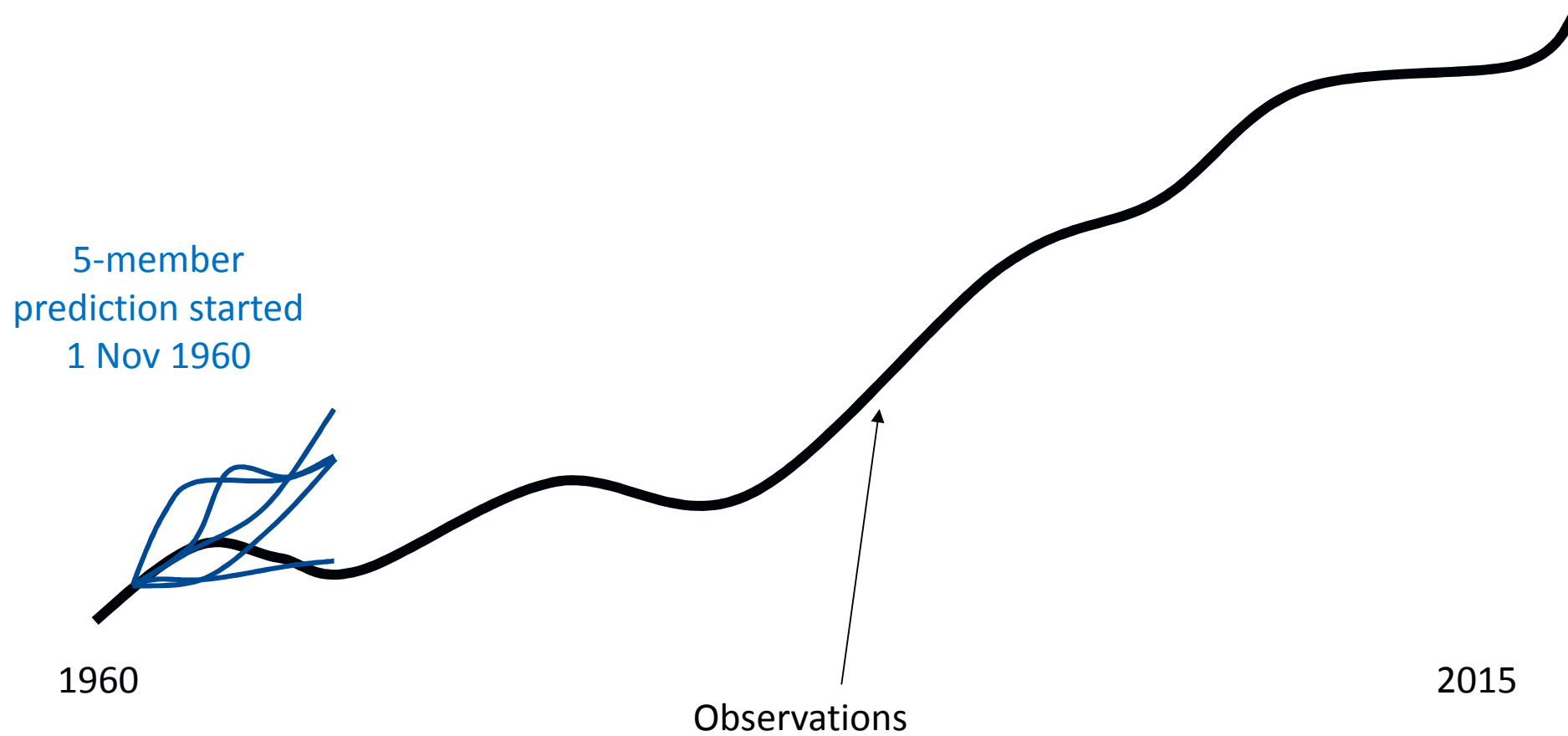
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# Decadal climate prediction



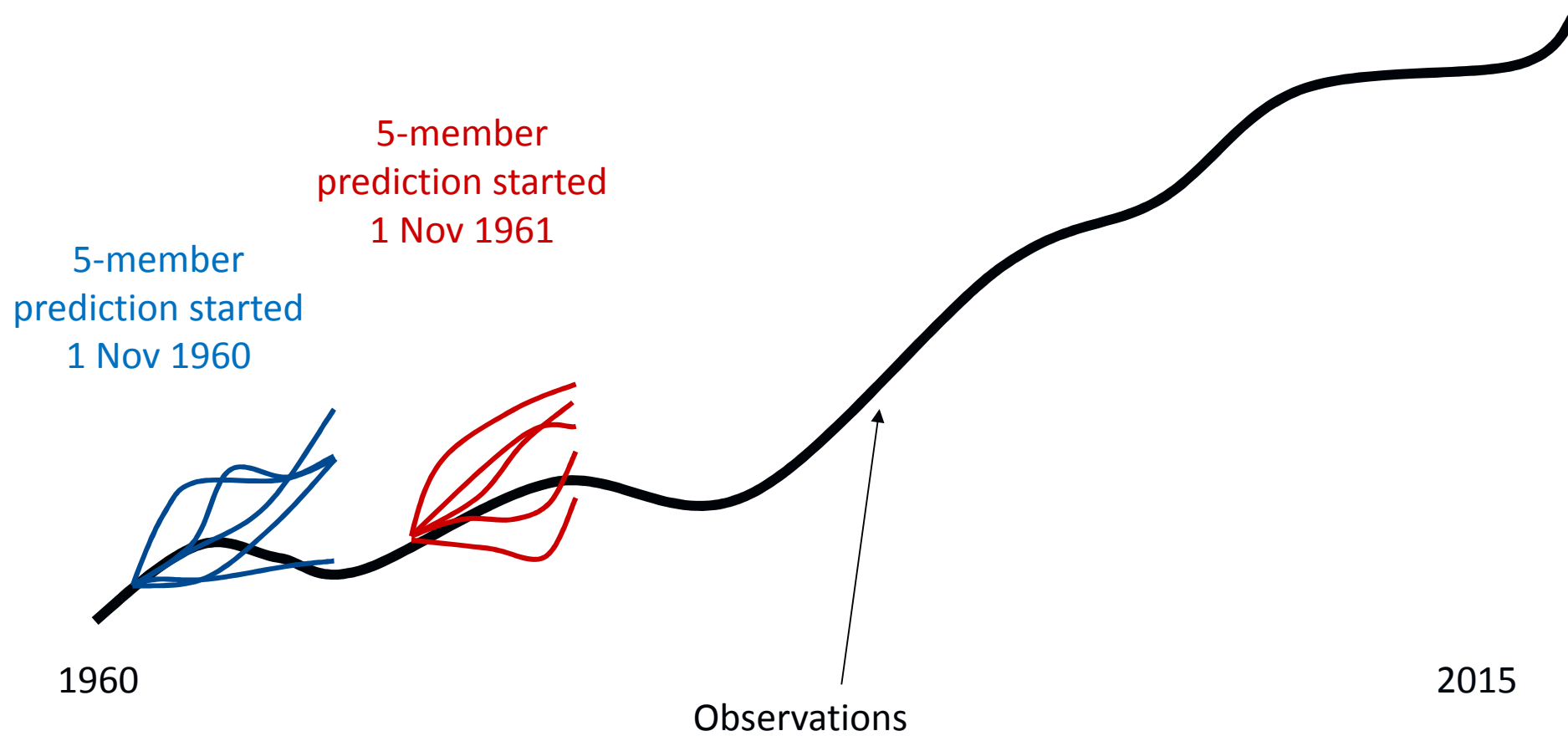
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# Decadal climate prediction



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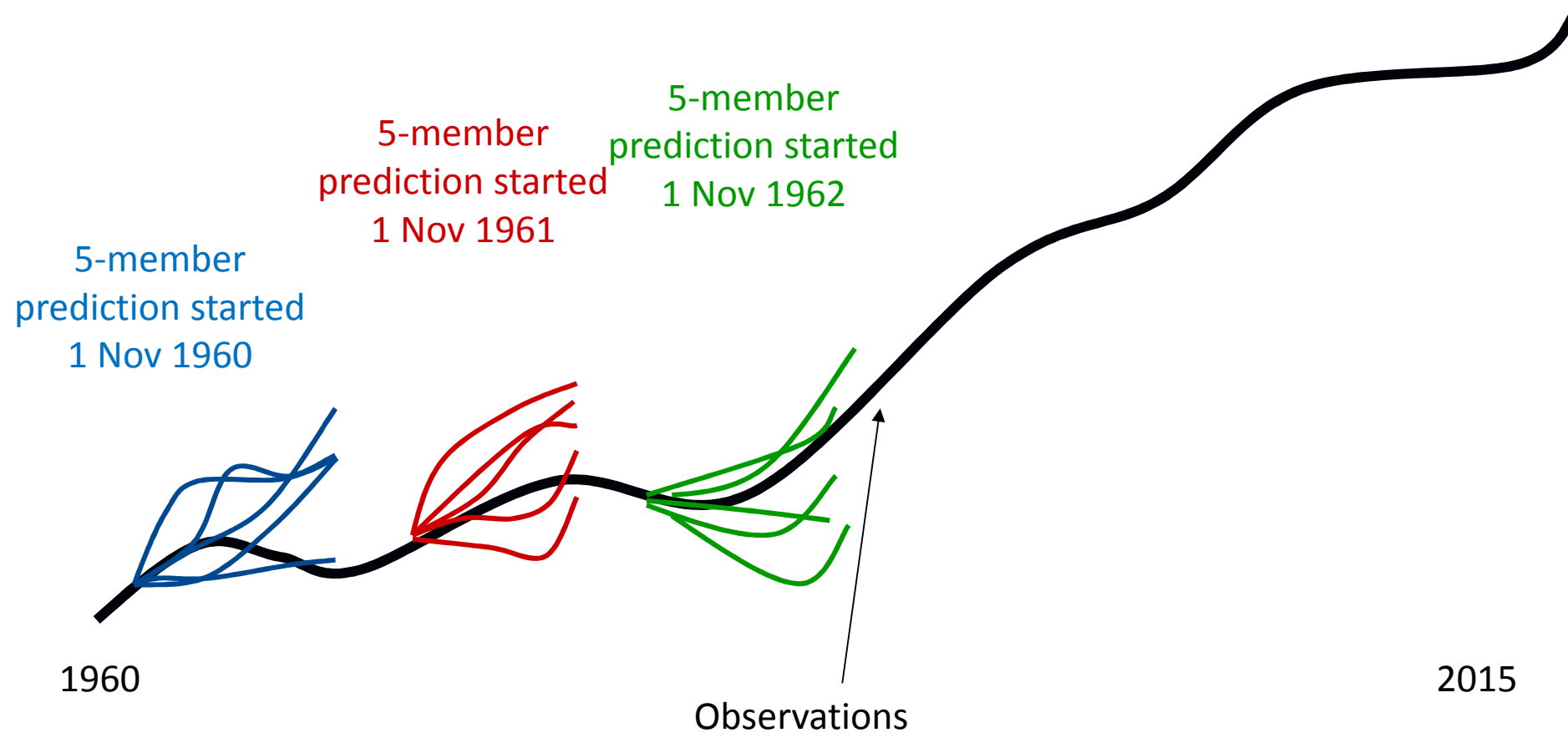




# Decadal climate prediction



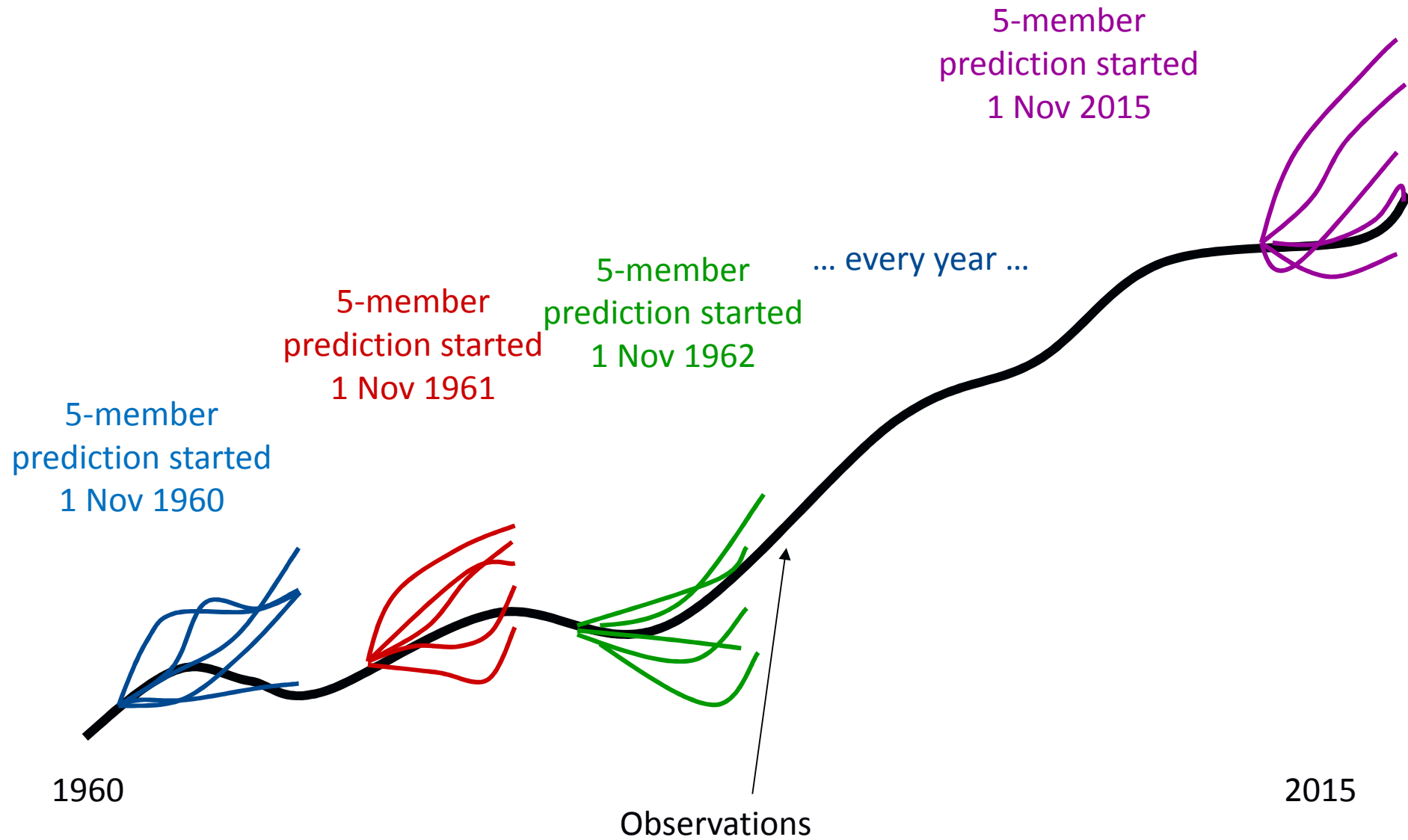
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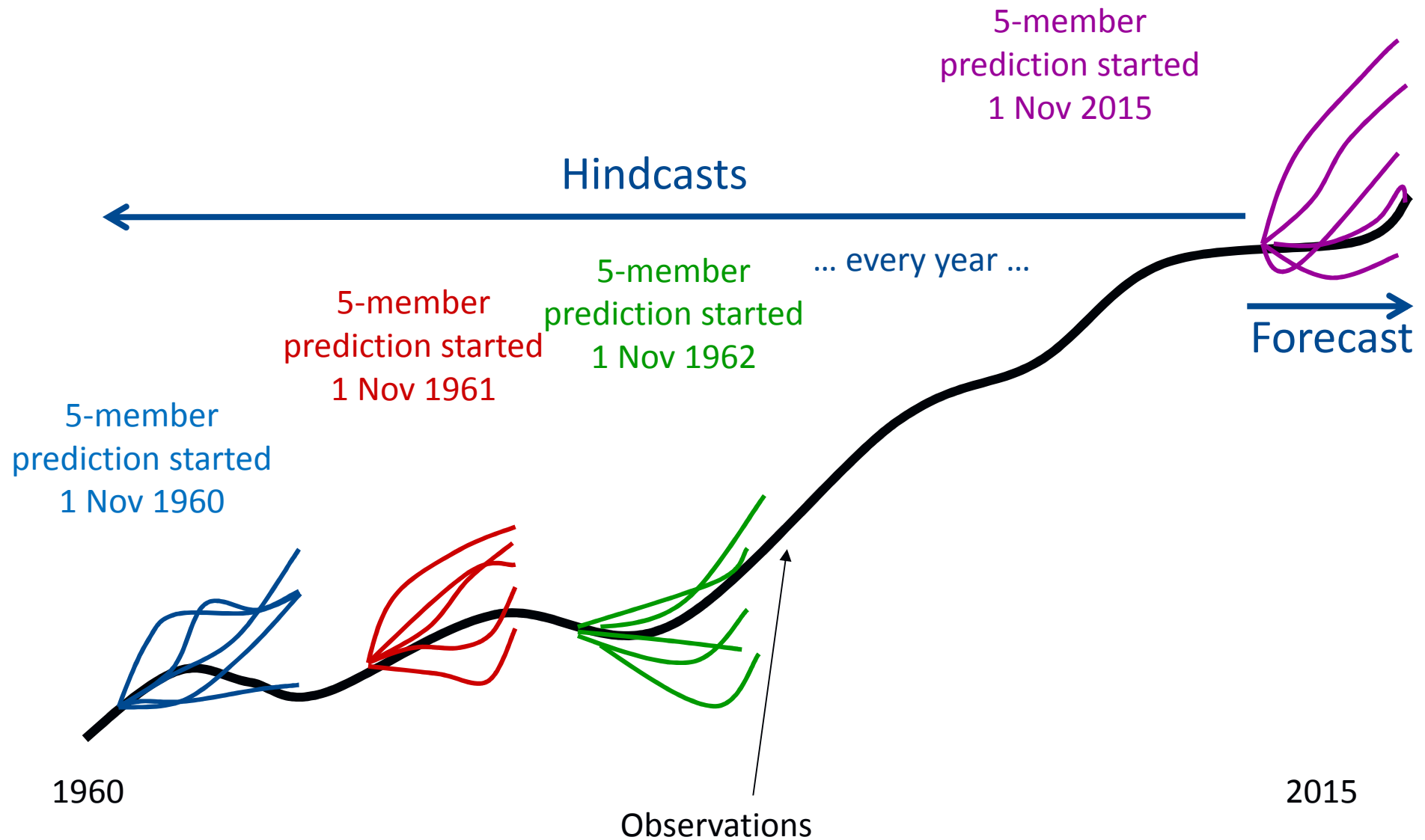


# Decadal climate prediction



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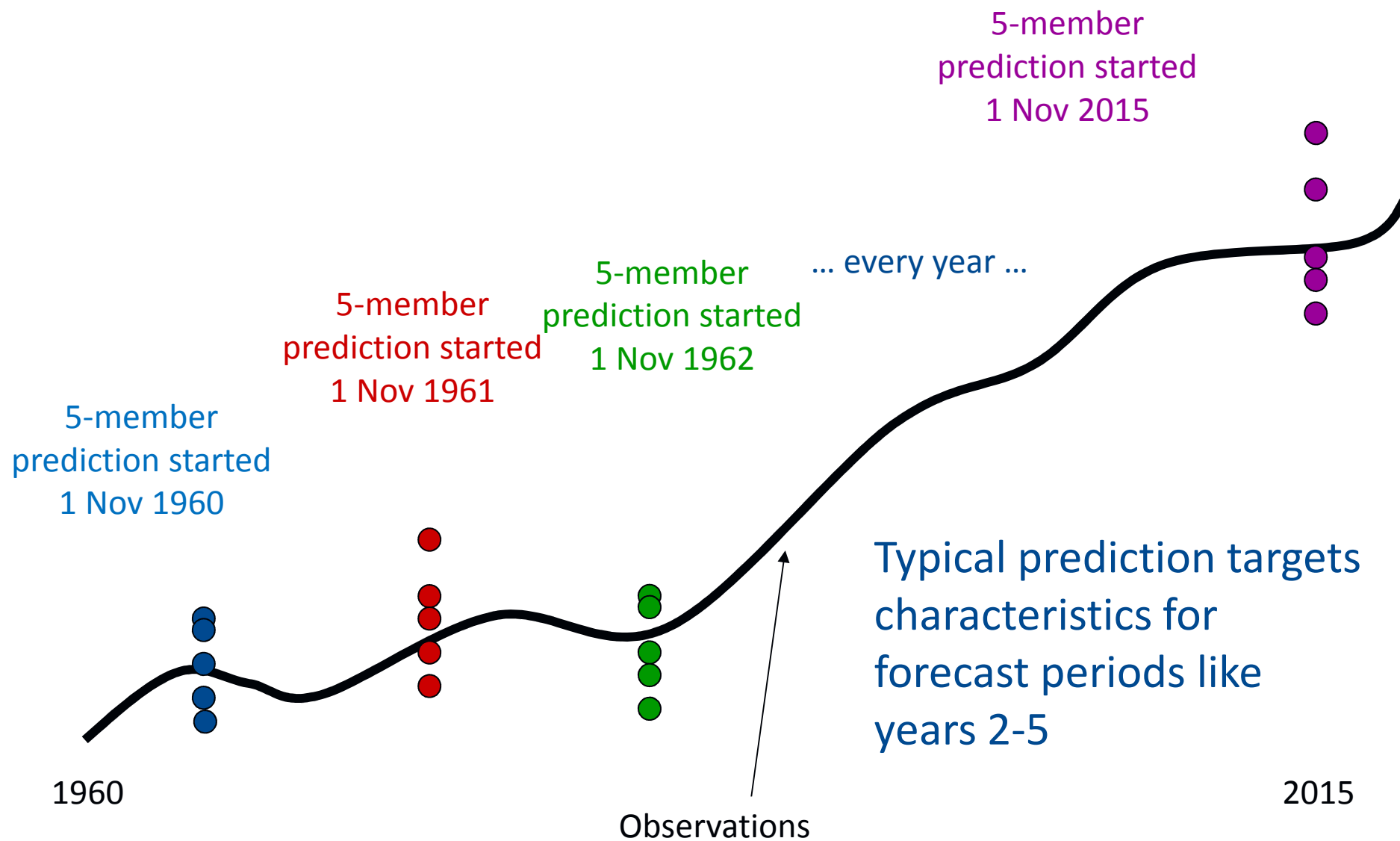




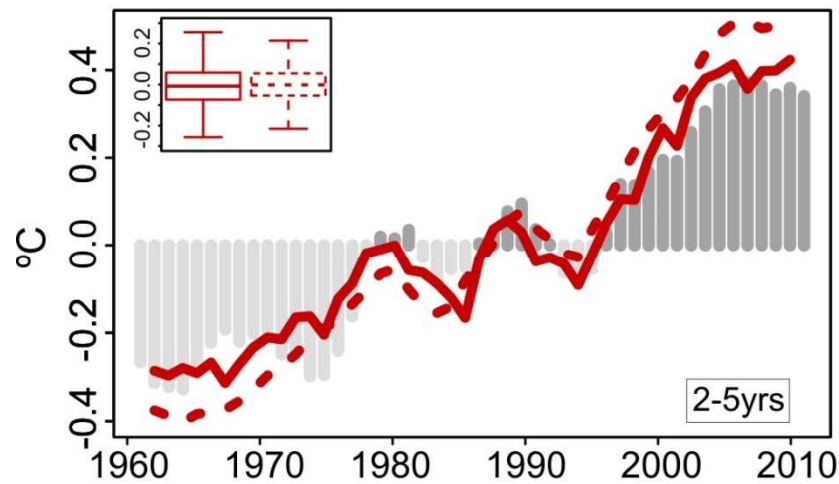
# Decadal climate prediction



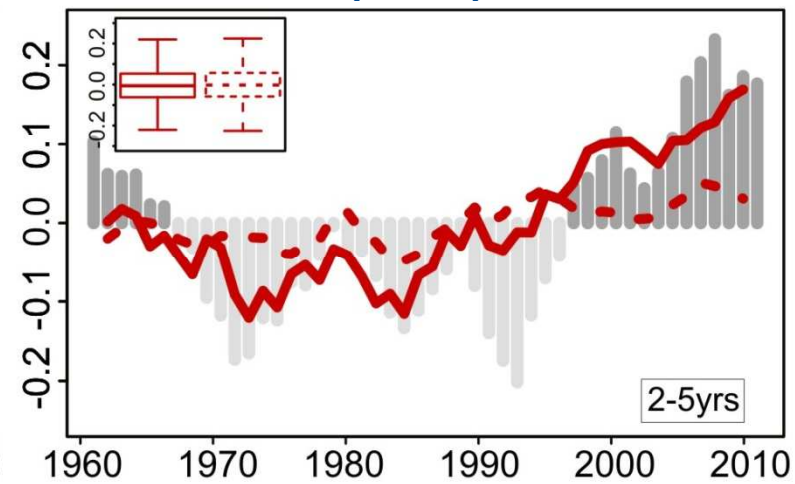
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## Global-mean surface temperature



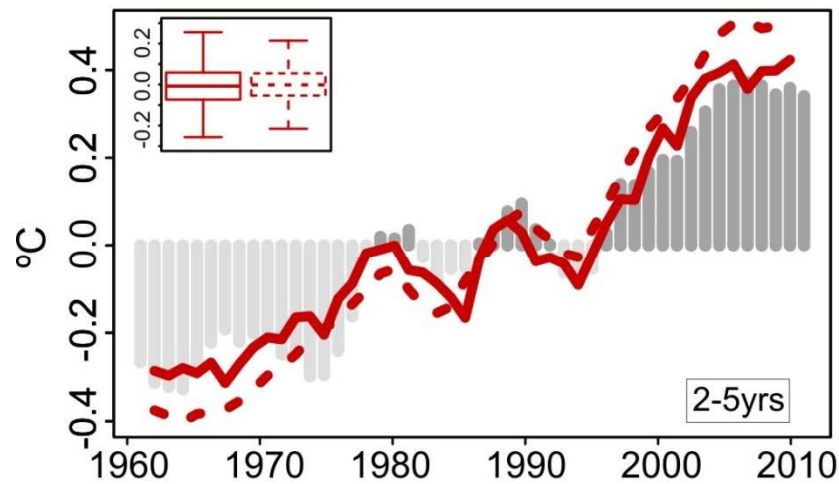
## Atlantic multi-decadal oscillation (AMO)



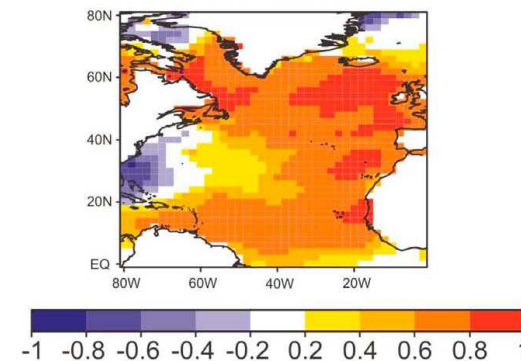
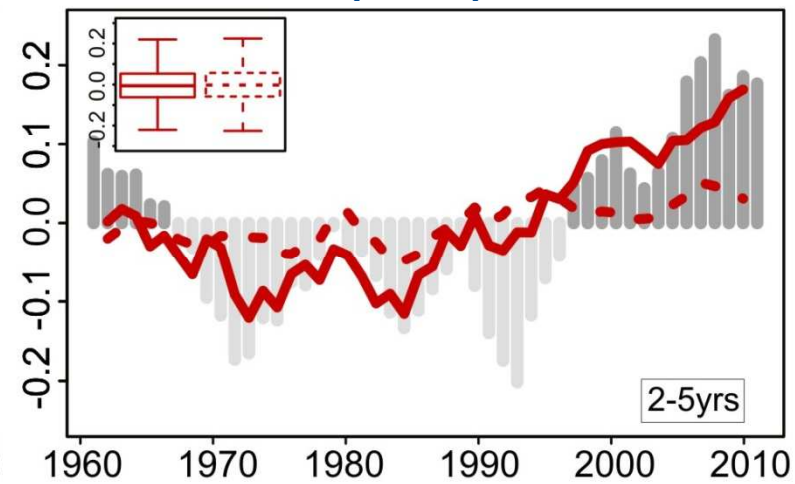
Doblas-Reyes et al. (2013, Nature Comms.)

— CMIP5 Init    - - - CMIP5 NoInit

## Global-mean surface temperature



## Atlantic multi-decadal oscillation (AMO)



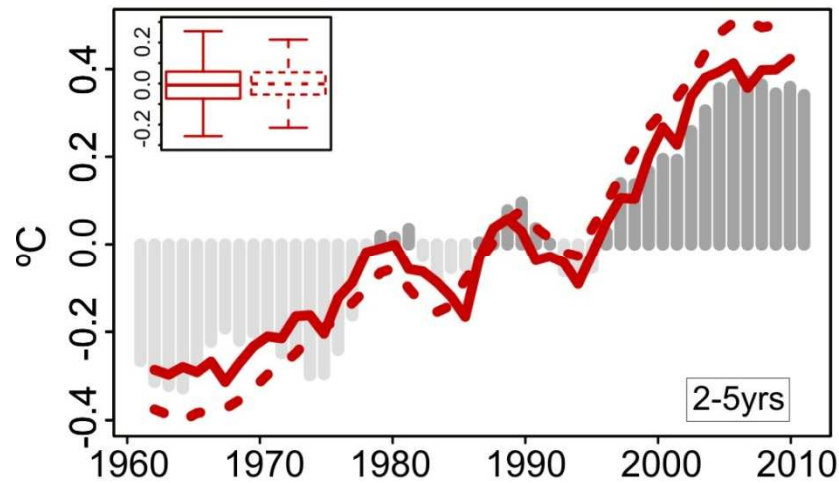
García-Serrano et al. (2012, GRL)

Doblas-Reyes et al. (2013, Nature Comms.)

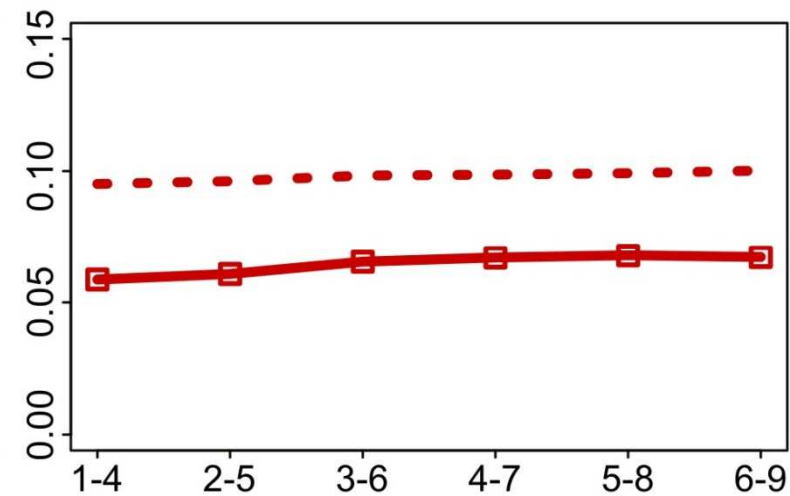
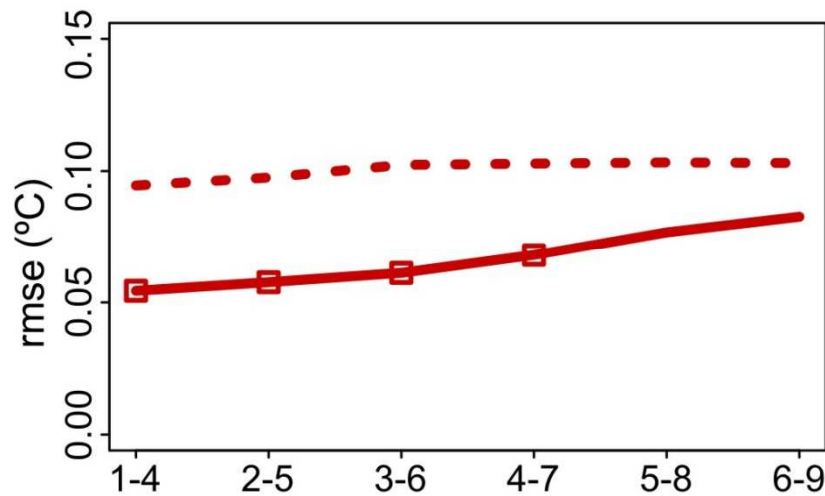
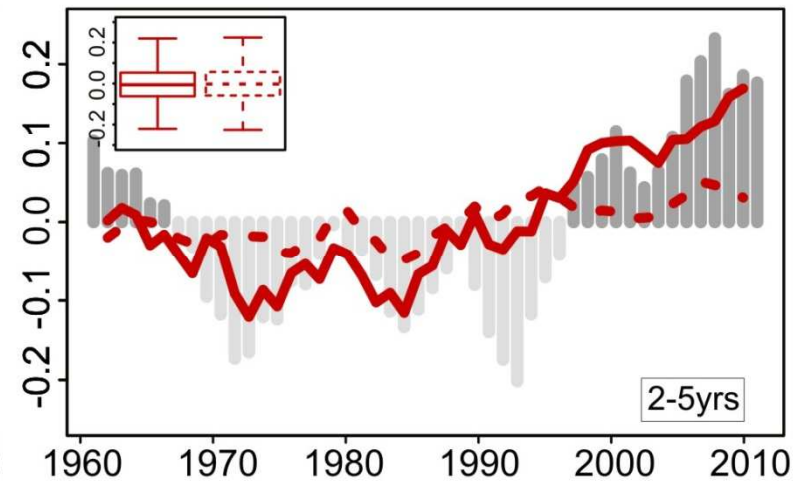




## Global-mean surface temperature



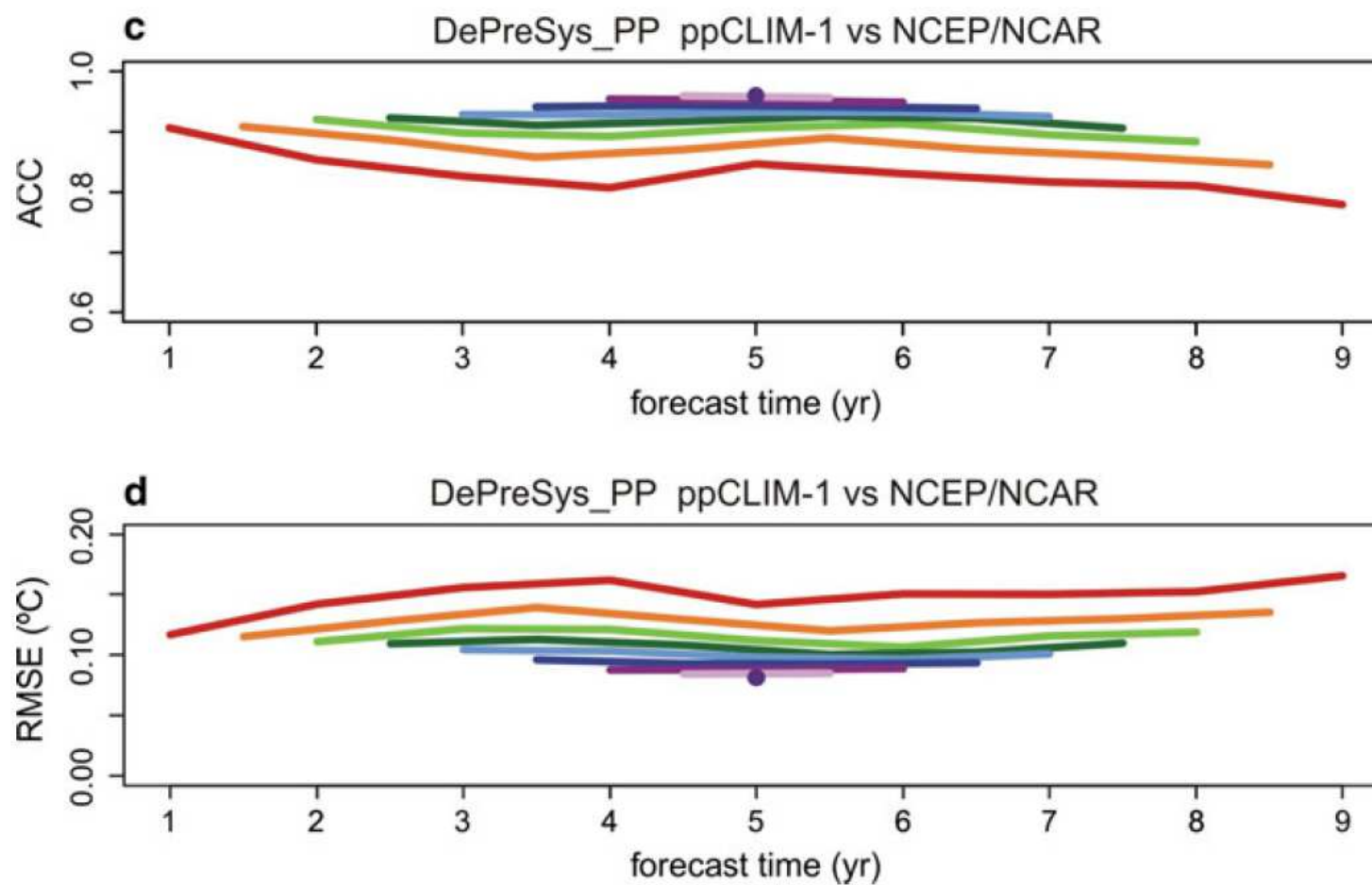
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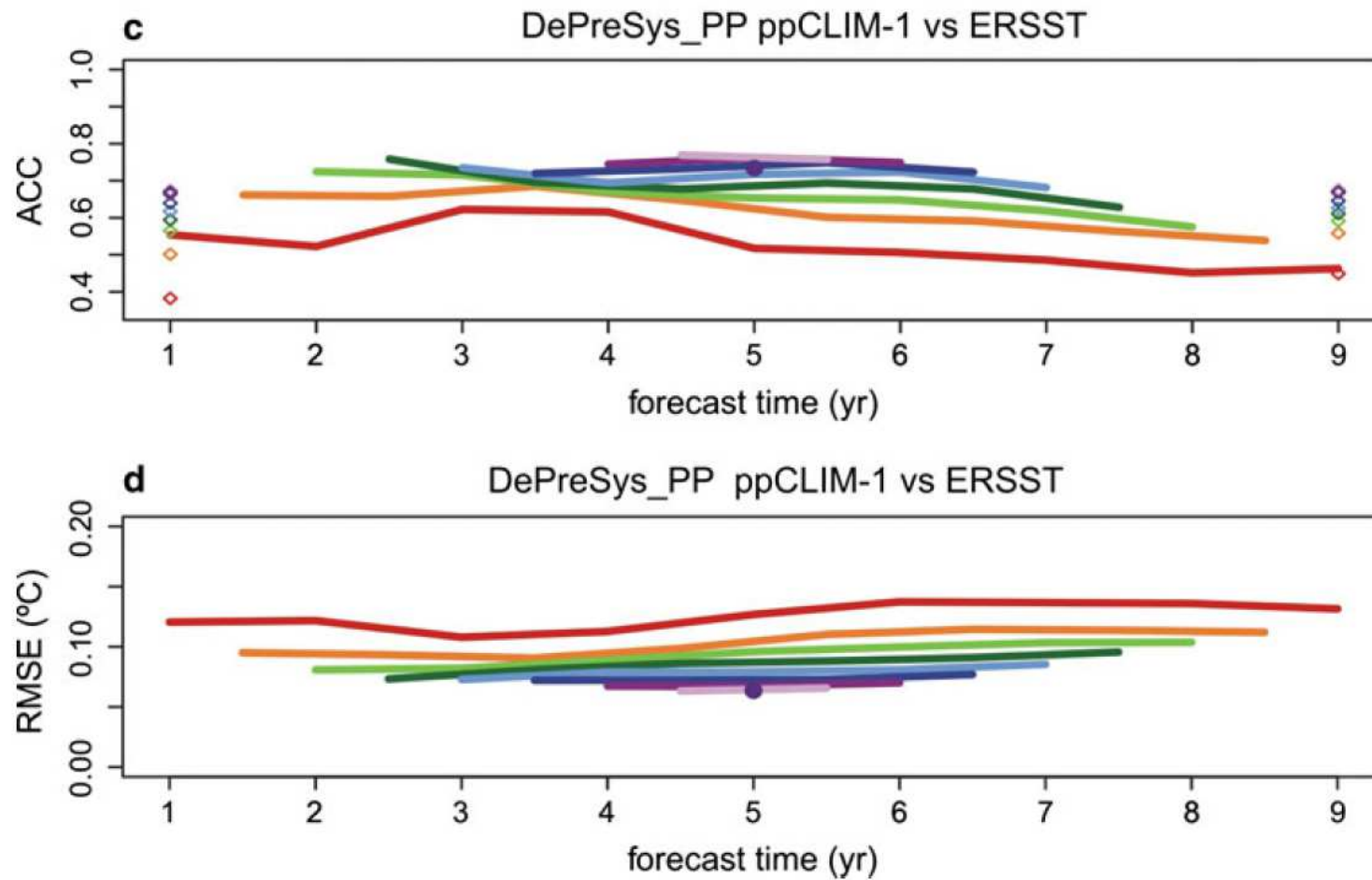
Doblas-Reyes et al. (2013, Nature Comms.)

— CMIP5 Init    - - - CMIP5 NoInit

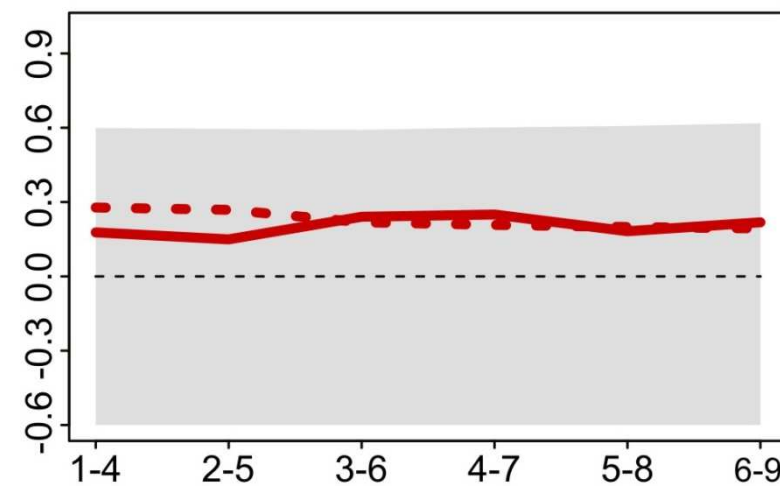
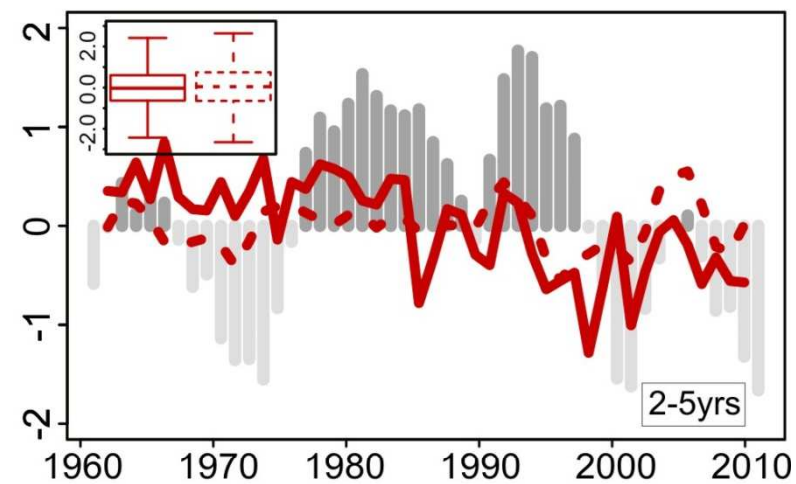
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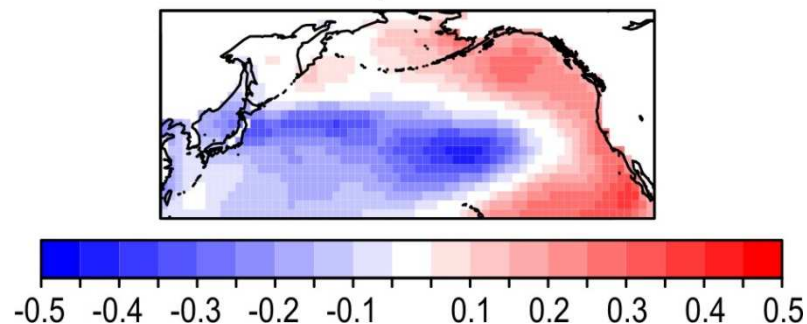
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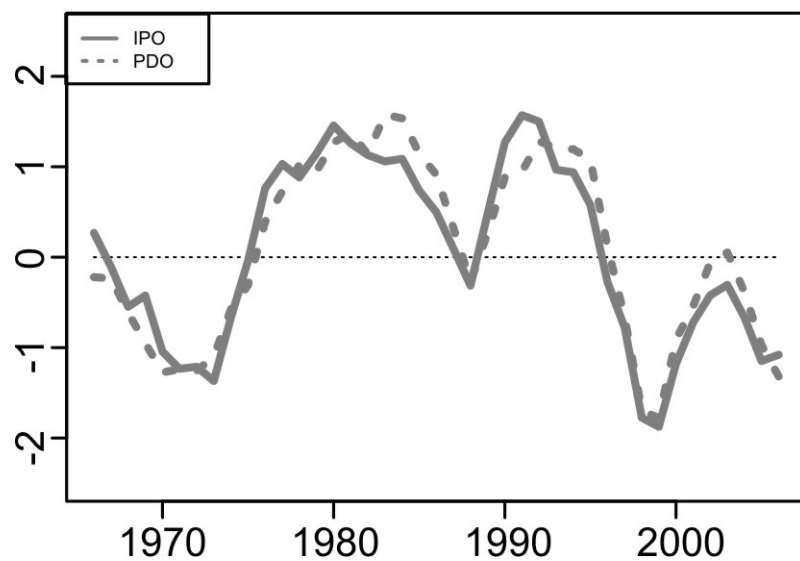
IPO



PDO pattern

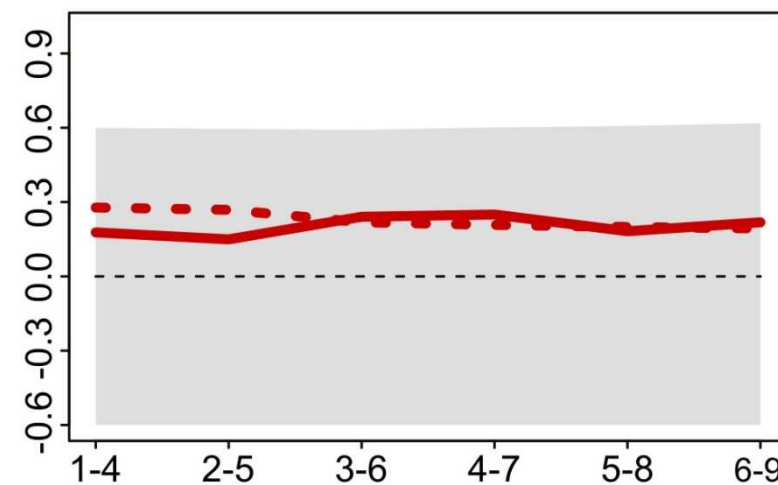
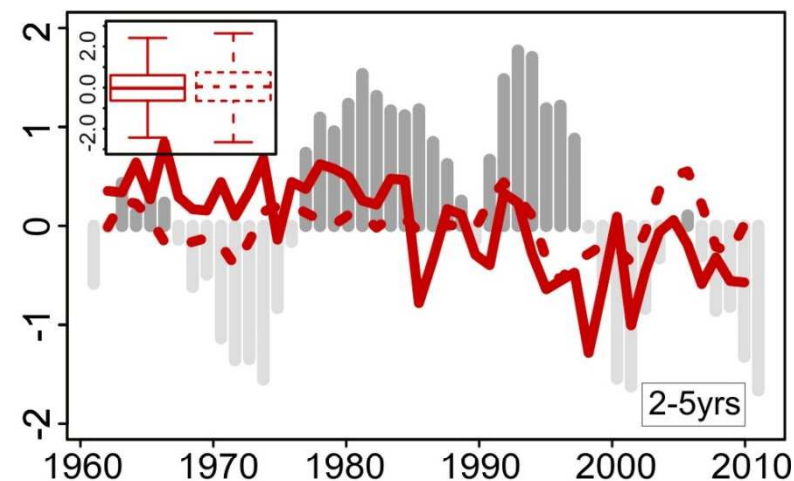


PDO/IPO 1966/69-2006/09



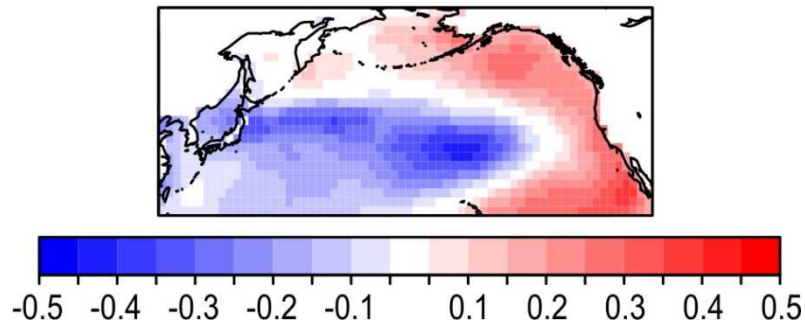
$r=0.96$

IPO

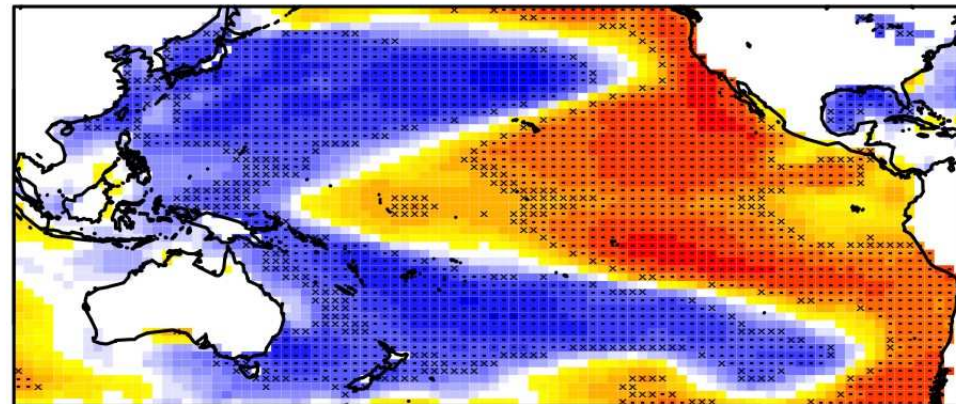




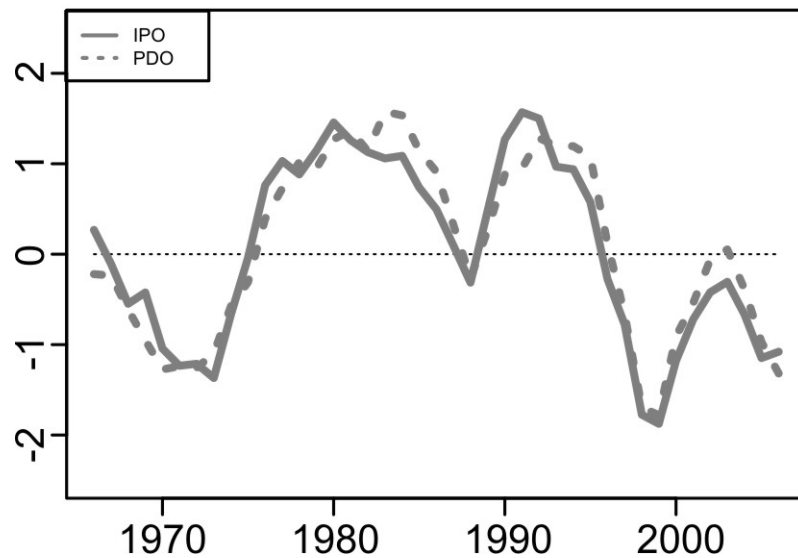
## PDO pattern



## PDO x SST

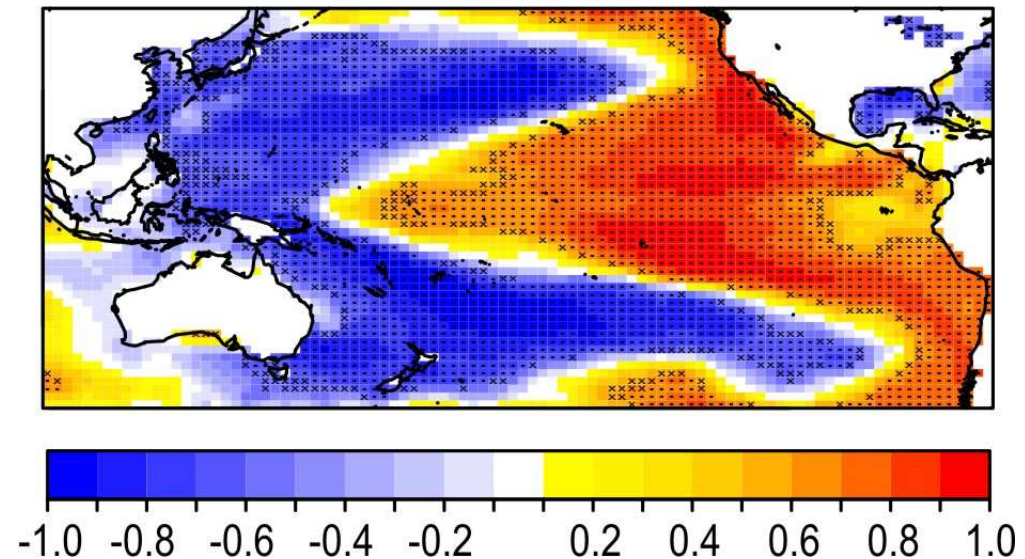


## PDO/IPO 1966/69-2006/09

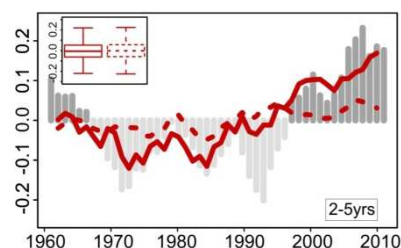


$r=0.96$

## IPO pattern



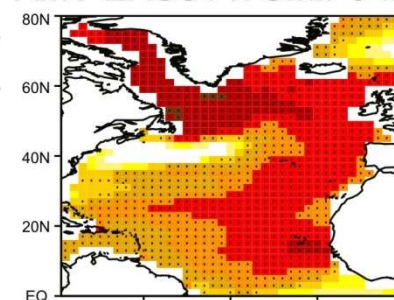
Correlation at each grid-point between multi-model ensemble-mean detrended SST and observed (ERSST) AMV index for INIT [left] and NoINIT [middle]



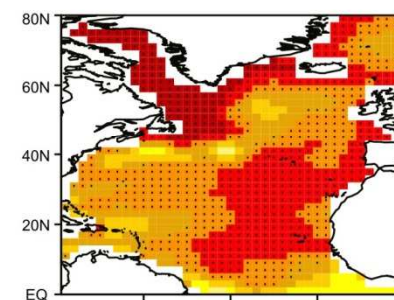
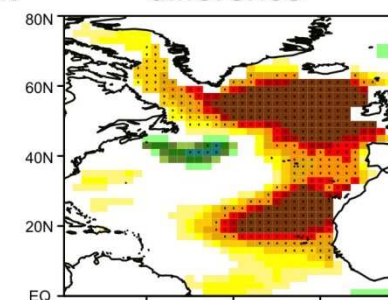
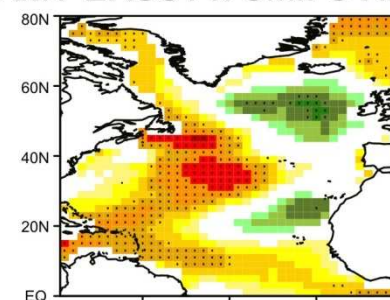
AMV-ERSST x CMIP5-Init

AMV-ERSST x CMIP5-NoInit

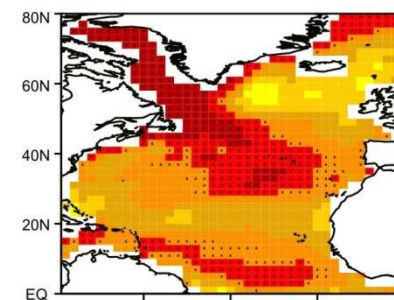
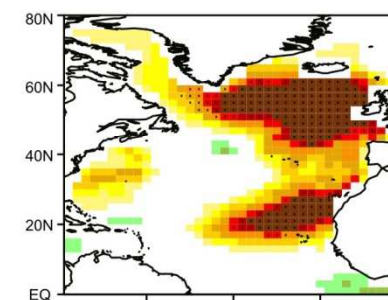
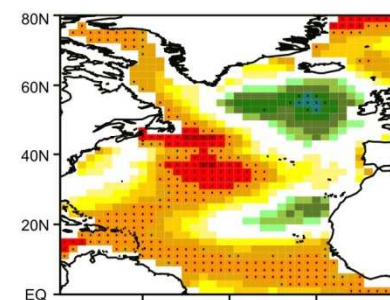
difference



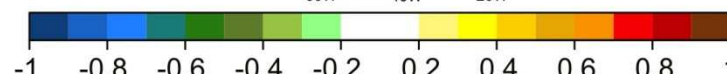
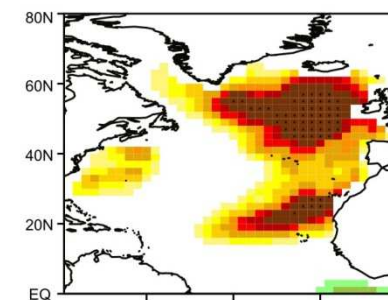
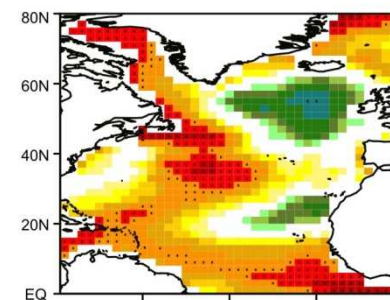
1-4yrs



3-6yrs

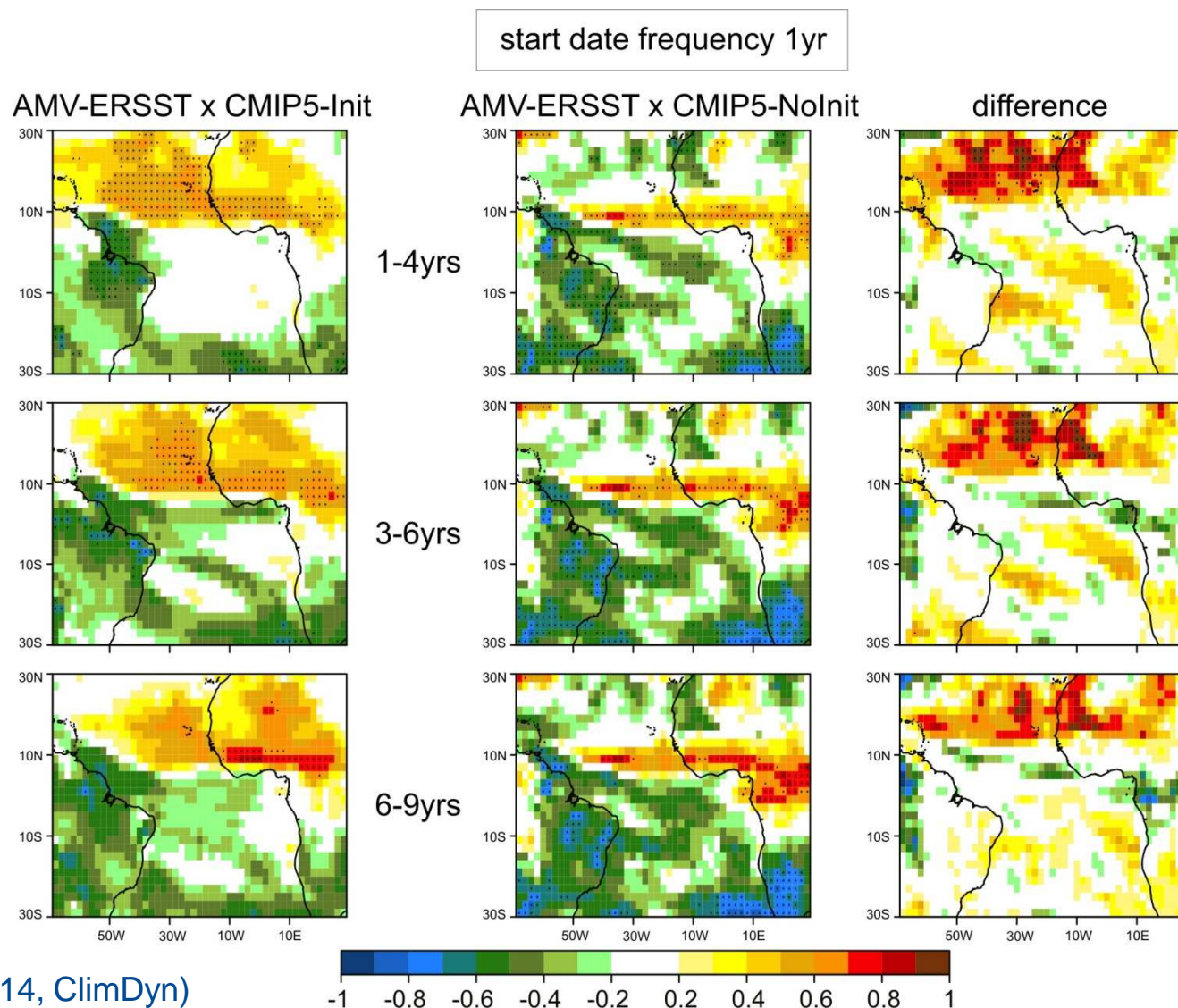
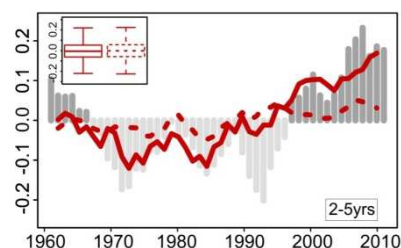


6-9yrs





Correlation at each grid-point between multi-model ensemble-mean precipitation and observed (ERSST) AMV index for INIT [left] and NoINIT [middle]



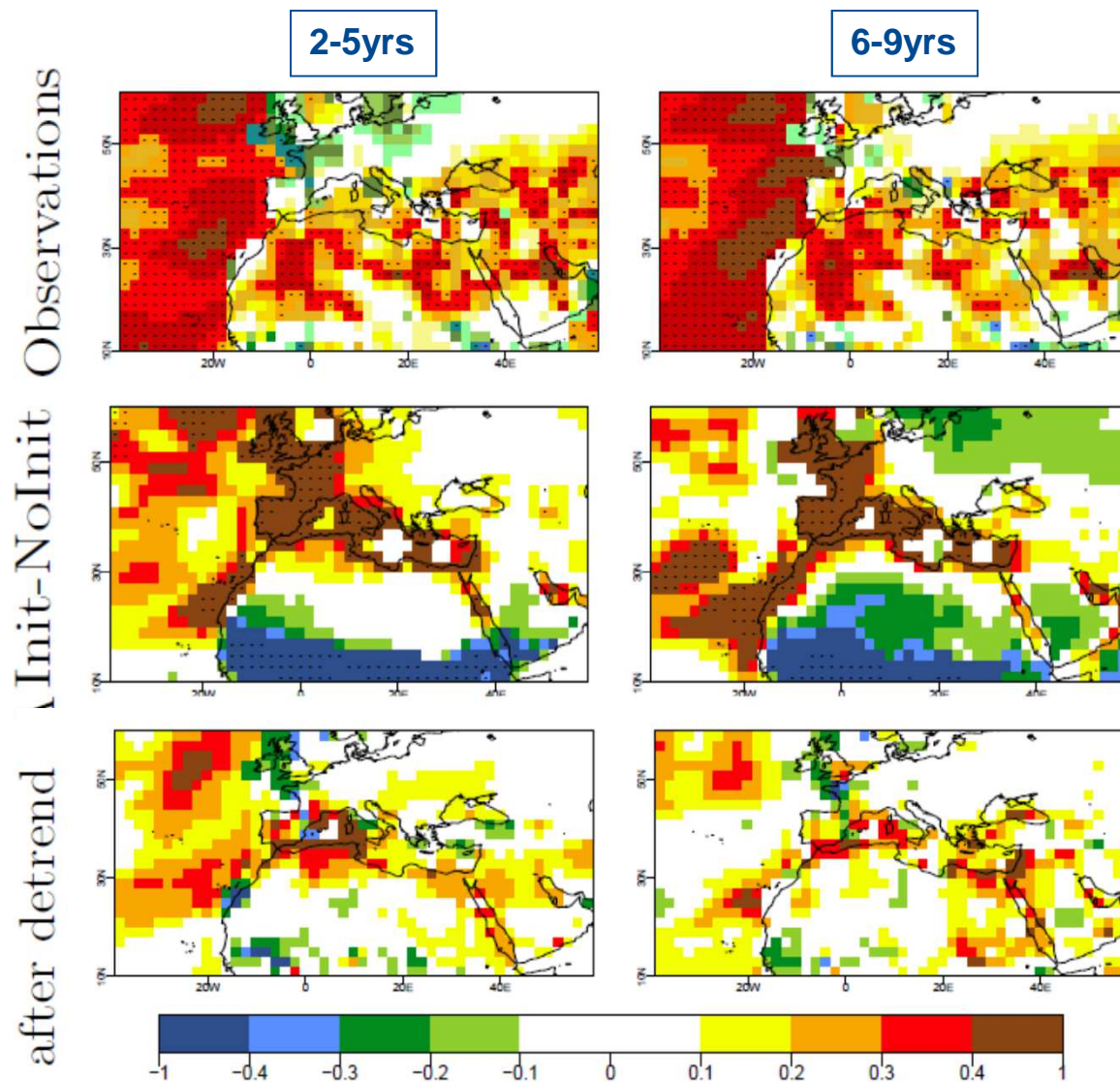
# Decadal climate prediction



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Correlation maps of observational [top] and multi-model ensemble-mean [middle] (INIT-NoINT) detrended surface temperature anomalies and the AMV index; [bottom] skill difference (INIT-NoINIT) of detrended surface temperature anomalies



Guemas et al. (2015)

- Formulate an appropriate and relevant scientific question
- Three components: hindcasts, forecasts, predictability
- Decadal prediction will benefit from being part of CMIP6:
  - better understanding (to, hopefully, reduce the drift) the systematic error
  - control runs for predictability estimates
- Other MIPs could benefit from the decadal-prediction MIP:
  - reduction of the systematic error by understanding the drift sources
  - continuous verification of the models
- Decadal prediction could be a very expensive part of CMIP
- Real-time decadal prediction exchange will continue and be enhanced (with more variables) whenever possible
- Contribute to climate services and WCRP grand challenges





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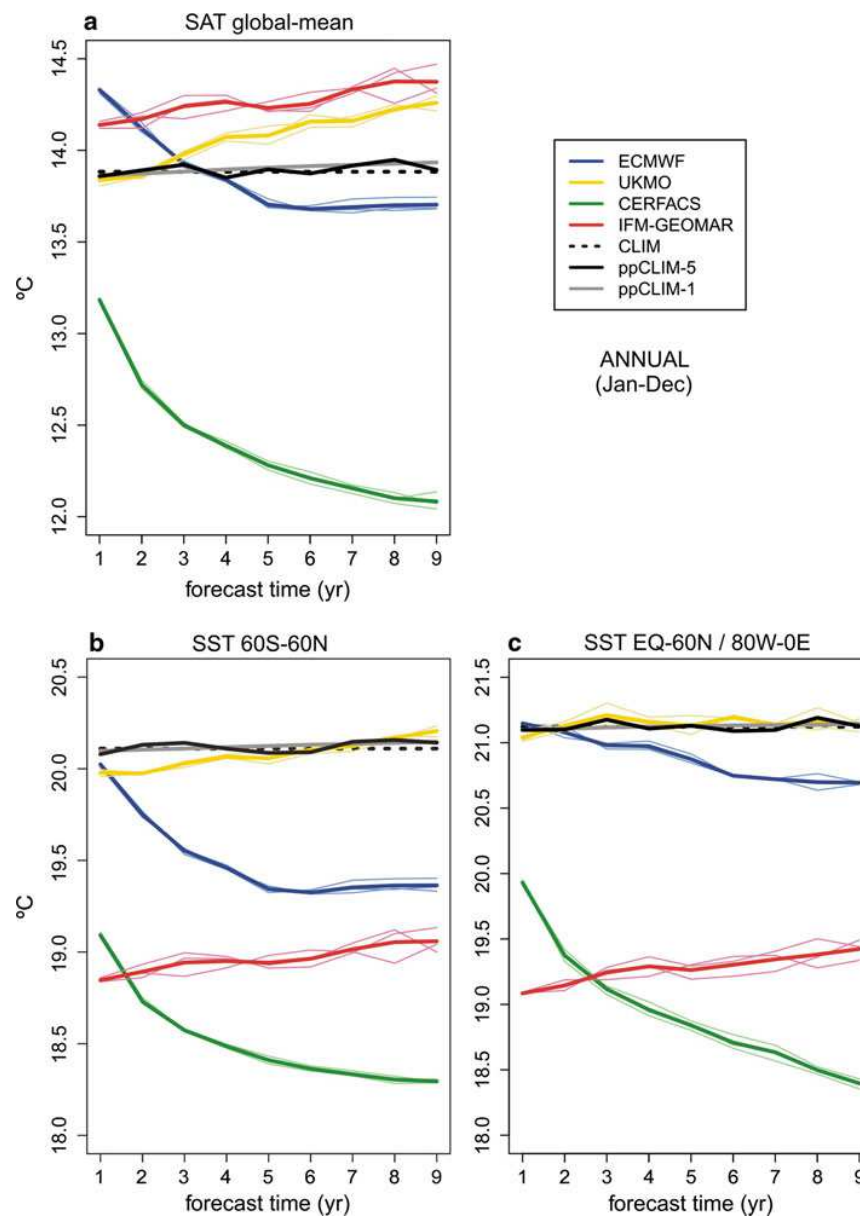
# *EXTRA SLIDES*

# Decadal climate prediction



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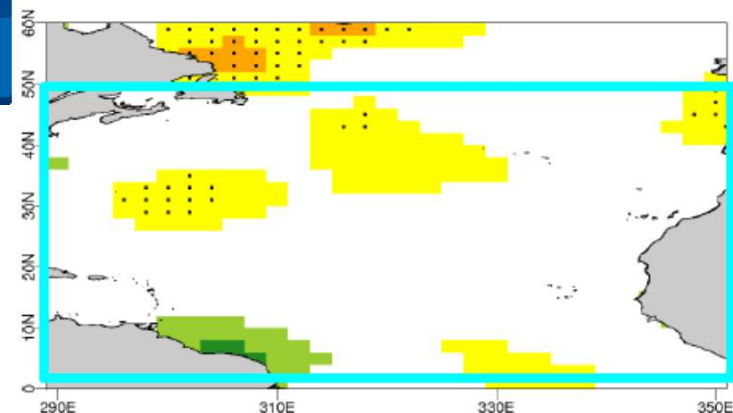
# Decadal climate prediction

GCMs	Initialized	Non-Initialized
GFDL CM2.1	10	10
HadCM3	10	10
MIROC5	6	3
MPI-ESM-LR	5	3

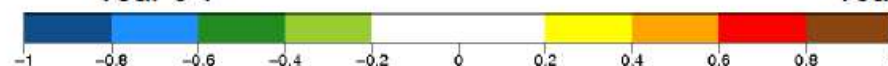
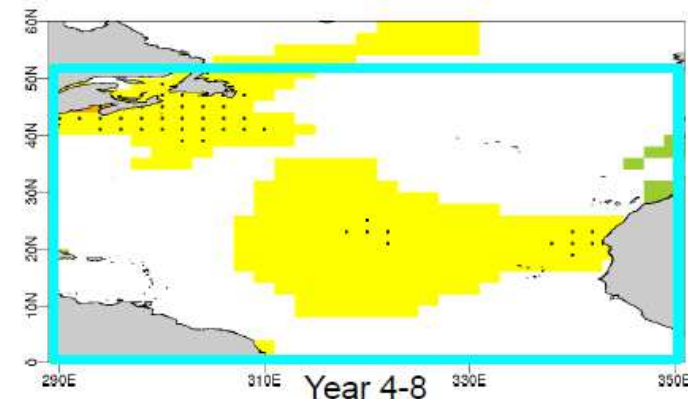
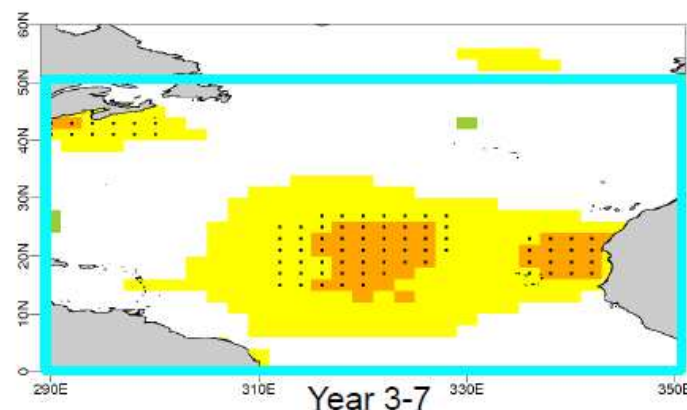
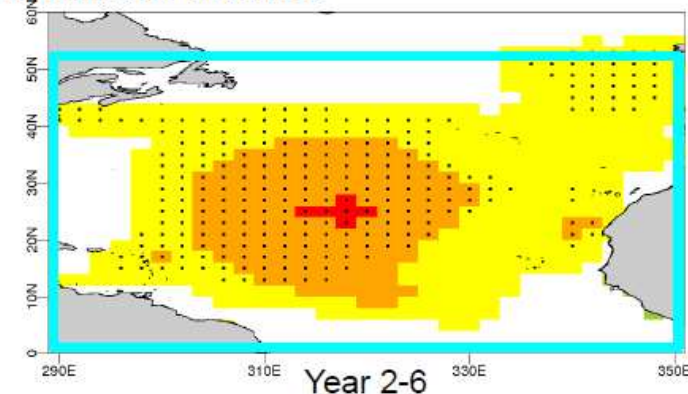
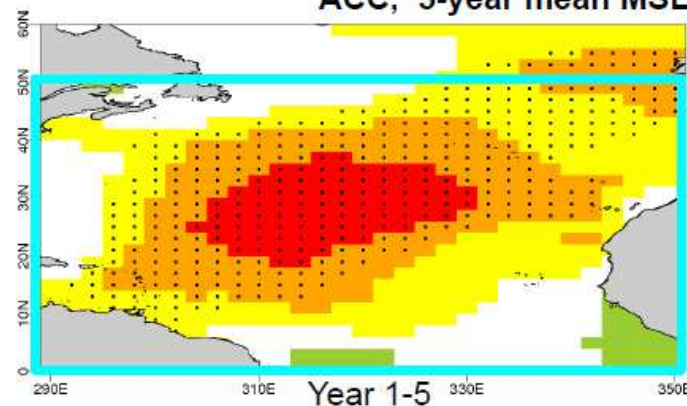
Start dates: yearly, 1961 to 2009

5-year mean predictions (1961-1966 to 2009-2013)

MSLP – Non-Initialized Forecasts



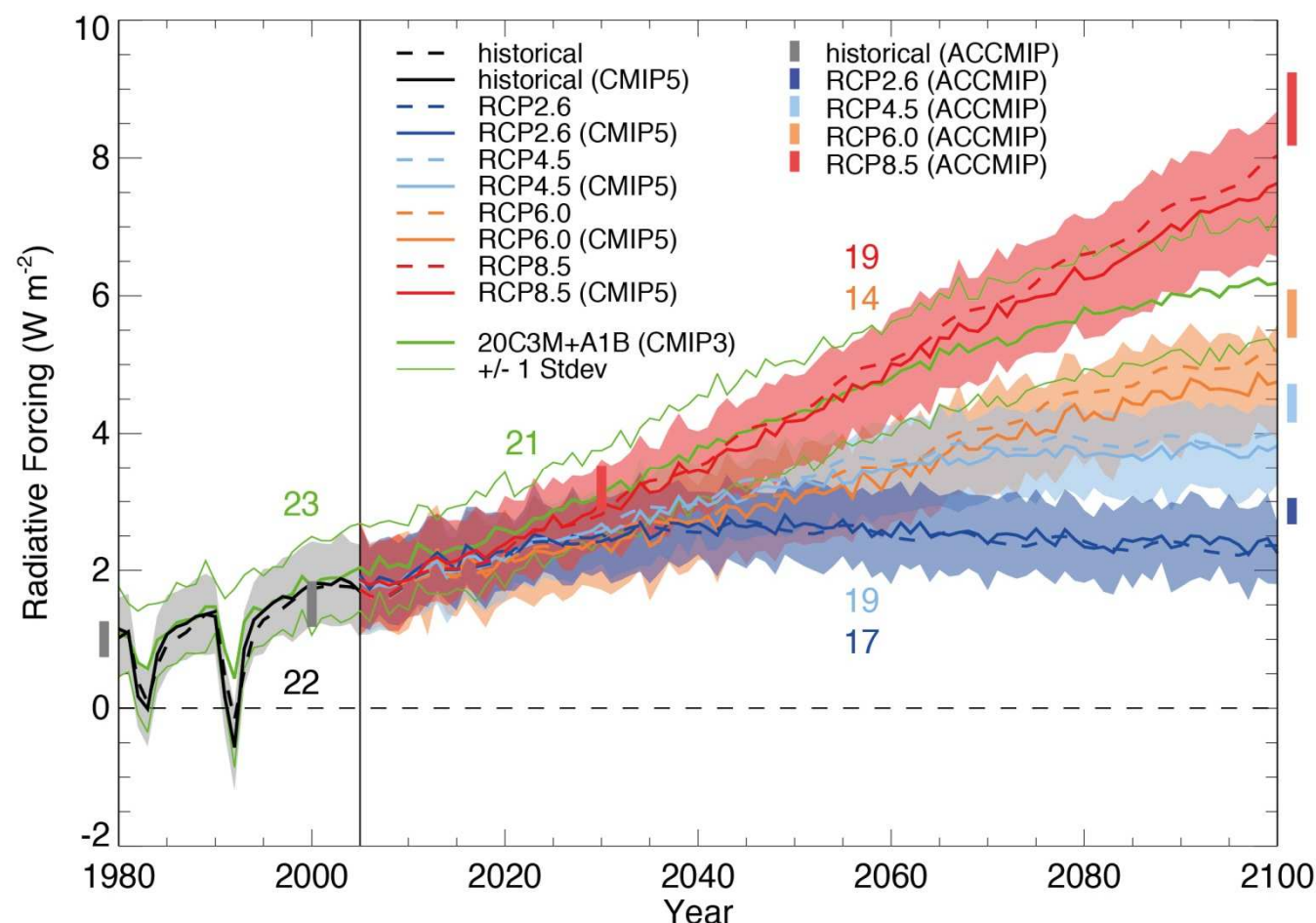
ACC, 5-year mean MSLP – Initialized Forecasts



Caron et al. (2015, GRL)

Global mean radiative forcing ( $\text{Wm}^{-2}$ , dashed) and effective radiative forcing (solid) between 1980 and 2100 with 1850 as baseline.

*There is little difference between the RCPs before 2040.*



## CMIP5 (initialized) hindcasts + historical&RCP4.5:

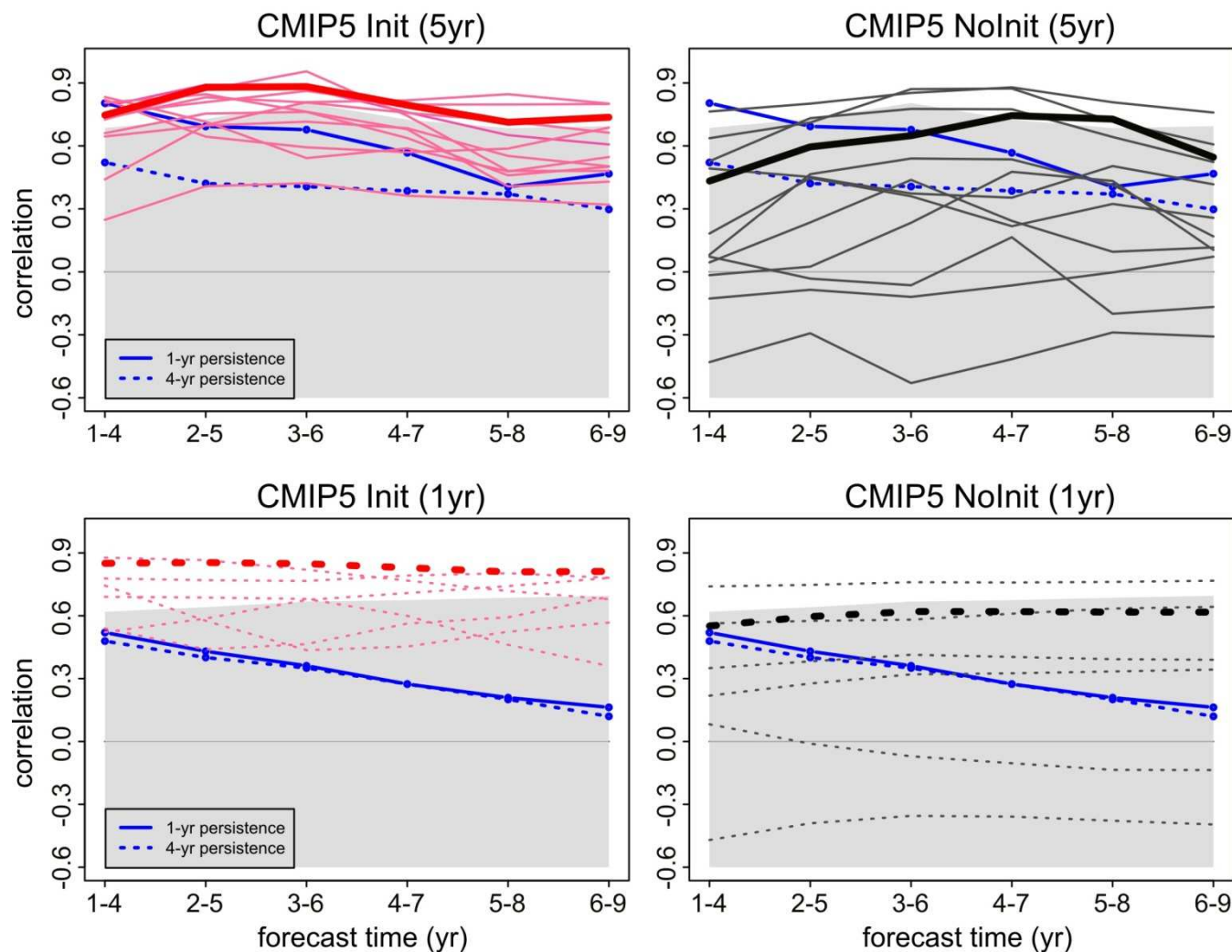
- 10-year long
- ensemble size & start date frequency: variable
- multi-model: 11 systems (5-yr) / 6 systems (1-yr)
- over the period 1960-2005

<b>forecast system</b>	<b>members</b>		<b>start date frequency</b>
	<i>Init</i>	<i>NoInit</i>	
HadCM3	10	10	5yr / 1yr
MRI-CGCM3	9	1	5yr
MIROC4h	3	3	5yr
MIROC5	6	1	5yr / 1yr
CanCM4	10	10	5yr / 1yr
CNRM-CM5	10	6	5yr
EC-EARTH2.3	10 / 5	11	5yr / 1yr
CMCC-CM	3	1	5yr
GFDL-CM2	10	10	5yr / 1yr
IPSL-CM5	6	6	5yr
ECHAM5/MPI-OM	10 / 3	10	5yr / 1yr



García-Serrano et al. (2014, ClimDyn)

## AMV forecast skill



Branstator and Teng (2012): inter-model differences in forced predictability larger than in initial-value predictability

Terray (2012): marked inter-model spread in the spatial response to anthropogenic forcing

- NEMO2: ORCA1 L46
- IFS [cy31r1]: T159 L62 (top 5hPa)
- LIM2: single ice category
- H-TESSEL – land-surface processes
- OASIS3 – coupler

## initial conditions

**atmosphere  
reanalysis  
(ERA-Interim)**

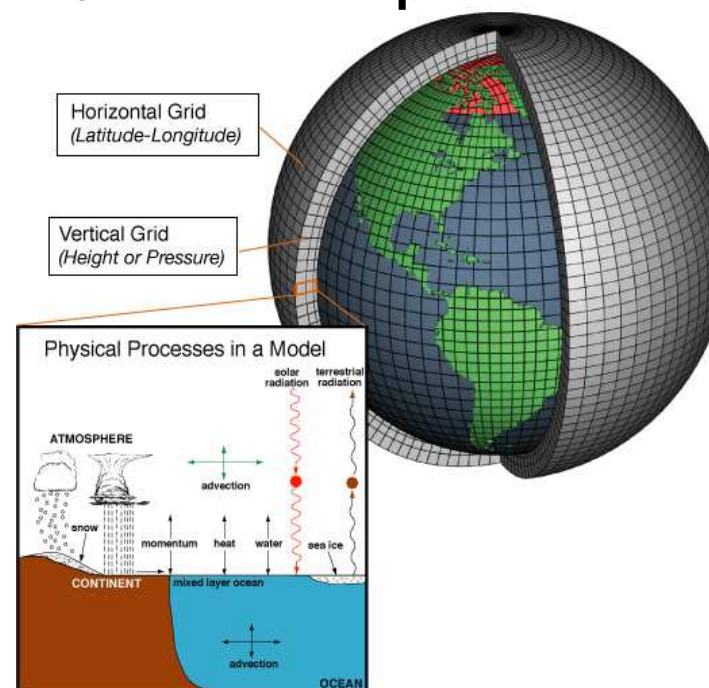
**ice  
reanalysis  
(IC3/BSC)**

**land reanalysis  
(ERA-Land)**

**ocean reanalysis  
(e.g. ORAS4)**



## EC-EARTH coupled model



- NEMO3.6: ORCA1 L75
- IFS [cy36r4]: T255 L91 (top 0.01hPa)
- LIM3: multiple (5) ice categories
- H-TESSEL – land-surface processes
- OASIS3-MCT – coupler

## initial conditions

**atmosphere  
reanalysis  
(ERA-Interim)**

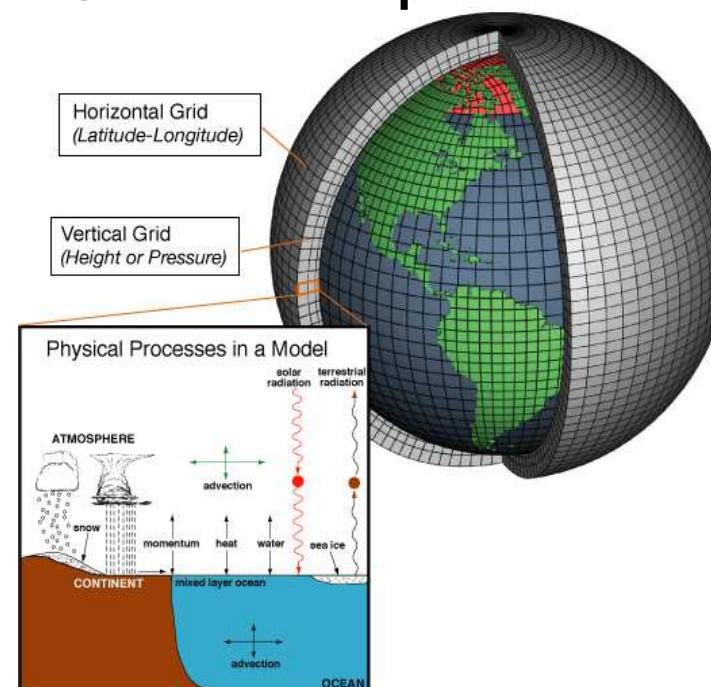
**ice  
reanalysis  
(IC3/BSC)**

**land reanalysis  
(ERA-Land)**

**ocean reanalysis  
(e.g. ORAS4)**



## EC-EARTH coupled model



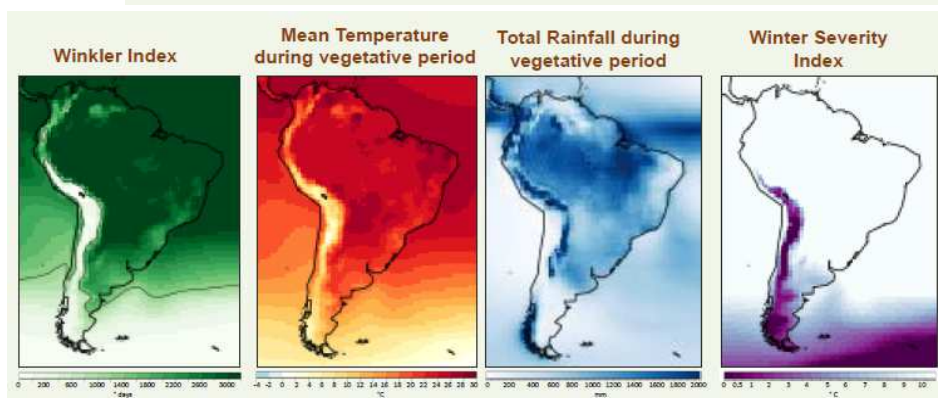


## Co-design of Wine Indices

In this study, four wine indices were defined in the co-design phase with the stakeholders (a wine company). Their main difference with the standard wine indices is that their definition is adapted to the Southern Hemisphere, where the warmest months are from October to April:

- The **Winter Severity Index** is the mean temperature of the coldest month (July in the Southern Hemisphere).
- The **Winkler Index** is the cumulated sum from October to April of all the degrees of daily mean temperature above 10°C.
- The **Mean Temperature during the vegetative period** is the average mean temperature from October to April.
- The **Total Rainfall during the vegetative period** is the total precipitation amount from October to April.

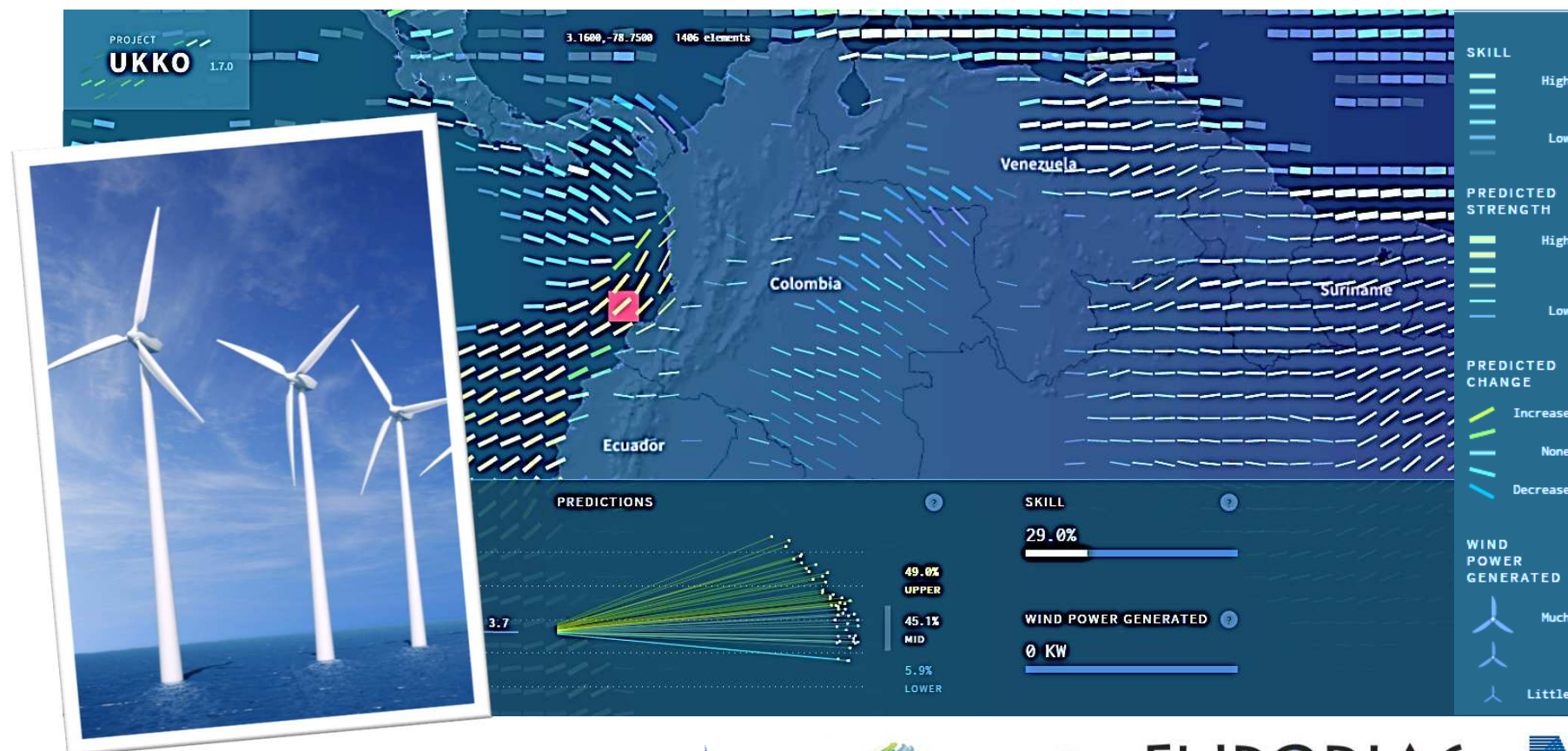
The selection is conditioned on the availability of climate models and observations in the study area. For example, in South America there are few observational datasets and few regional models available. In this cases, it is preferable to employ daily reanalysis datasets and global climate models instead (in this study, respectively the ERA-Interim and the CMIP5 models).



# Decadal climate prediction → 'climate services'



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